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Proceedings Paper:

Shanker, Sanil, Turner, Aaron, Sherly, Elizabeth et al. (1 more author) (2010) Sequential Data Mining using Correlation Matrix Memory. In: Proceedings. ICNIT 2010 International Conference on Networking and Information Technology. Proceedings. ICNIT 2010 International Conference on Networking and Information Technology, 11-12 Jun 2010 IEEE, PHL, pp. 470-472.

https://doi.org/10.1109/ICNIT.2010.5508469

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Sequential Data Mining using Correlation Matrix Memory

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Abstract: This paper proposes a method for sequential data mining using correlation matrix memory. Here, we use the concept of the Logical Match to mine the indices of the sequential pattern. We demonstrate the uniqueness of the method with both the artificial and the real datum taken from NCBI databank.

Keywords: Correlation matrix memory, Logical Match, Sequential data mining.

I INTRODUCTION

In this paper we put forward a unique method of search for sequential patterns using correlation matrix memory by Logical Match. The correlation matrix memory is a particular type of binary associative neural network (Austin, 1994). In Logical Matching strategy, the sequence is arranged so that each element in the pattern coincides with it's corresponding index and then proceed to match logically the indices of the subsequence with those of the text pattern (Sanil et al,2010). A correlation matrix memory can be represented as a m X n binary matrix D, where m and n are input width and output width respectively. For a given sequential pattern the input binary vector I_{SP} operates logical AND with reference matrix (database D) gives I_{SP}^{T} . During the transformation, $O_P \leftarrow I_P$.D, we mine the indices of both the pattern and text using logical matching. where $I_{SP}^{T} = O_{SP}$, the output binary vector of the sequential pattern.

II METHOD

1.Initialize Text and Pattern

2.Initialize input binary vector $I_P \leftarrow$ Pattern and $I_T \leftarrow$ Text

3.Create the reference database D

4. Select (indices of Pattern) when $O_P \leftarrow I_P \cdot D$

5. Select (indices of Text) when $O_T \leftarrow I_T \cdot D$

6. Match (indices of Pattern) with (indices of Text) in the increasing order of indices.

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III SIMULATION WITH ARTIFICIAL DATUM

Text => <CTCACTCCTC>

Pattern => <CTC>

Initialize $\{0001 \leftarrow A, 0010 \leftarrow T, 0100 \leftarrow G, 1000 \leftarrow C\}$

Shift the Text(Table 1) so that the input binary vector I_{Text}

operates logical AND with reference matrix (Database D)

gives I_{Text}^{T} (Table 2).

Input Text (I_{Text})

Database D

| C T C A C T C C | С | Т | С | Α | С | Т | С | С | Т |
|-----------------|---|---|---|---|---|---|---|---|---|
|-----------------|---|---|---|---|---|---|---|---|---|

C AND 0/1 0/1 0/1 0/1

| 1 | 0 | 1 | 0 | 1 | 0 | 1 | 1 | 0 | 1 | • | 0 | 0 | 0 | 1 |
|---|----------|---|---|---|---|---|---|---|---|---|---|---|---|---|
| 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | • | 0 | 0 | 1 | 0 |
| 0 | 1 | 0 | 0 | 0 | 1 | 0 | 0 | 1 | 0 | • | 0 | 1 | 0 | 0 |
| 0 | 0 | 0 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | • | 1 | 0 | 0 | 0 |
| | Table 1. | | | | | | | | | | | | | |

$$\checkmark \downarrow \downarrow \downarrow$$

x y z

| Indices | w | х | У | Z |
|---------|---|---|---|---|
| 10 | 0 | 0 | 0 | 1 |
| 9 | 0 | 1 | 0 | 0 |

w

| 8 | 0 | 0 | 0 | 1 |
|---|---|---|---|---|
| 7 | 1 | 0 | 0 | 0 |
| 6 | 0 | 0 | 0 | 1 |
| 5 | 0 | 1 | 0 | 0 |
| 4 | 0 | 0 | 0 | 1 |
| 3 | 0 | 0 | 0 | 1 |
| 2 | 0 | 1 | 0 | 0 |
| 1 | 0 | 0 | 0 | 1 |

That is, indices of the Text, < w(7);

x(2,5,9);

z(1,3,4,6,8,10)>

Shift the Pattern (Table 3) so that the input binary vector

I_{Pattern} operates logical AND with reference matrix

(Database D) gives $I_{Pattern}^{T}$ (Table 4).

Input Pattern (I_{Pattern})

Database D

| C 0/1 | | | (| C | | AND | | 0/1 | | 0/1 | |
|----------|------|--------------|--------|------|-----|-------|-------|------|-----|---------|--------|
| | 1 | 0 | 1 | | • | 0 | 0 | 0 | | 1 |] |
| | 0 | 0 | 0 | | • | 0 | 0 | 1 | | 0 | |
| | 0 | 1 | 0 | | • | 0 | 1 | 0 | | 0 | |
| | 0 | 0 | 0 | | • | 1 | 0 | 0 | | 0 | |
| L | | | | Tal | ble | 3. | | | | | 」↓ |
| ↓ | | \downarrow | | | Ļ | | | | | | · |
| w | | | x | | 3 | 7 | | z | | | |
| | | Tabl | e 4. 🗍 | Fran | spc | se of | Input | Patt | ern | IPatter | T 1 |
| | Indi | ces | w | | x | | у | | z | | |

| 3 | 0 | 0 | 0 | 1 |
|---|---|---|---|---|
| 2 | 0 | 1 | 0 | 0 |
| 1 | 0 | 0 | 0 | 1 |

That is, indices of the pattern, <x(2);

Locating Pattern in the Text:

z: <u>1</u> 3 <u>4</u> 6 <u>8</u> 10

x: <u>2</u> <u>5</u> <u>9</u>

z: 1 <u>3</u> 4 <u>6</u> 8 <u>10</u>

Here, Pattern CTC is repeating in the locations (1,2,3);(4,5,6);(8,9,10) of the text.

IV EXPERIMENTAL RESULTS

For testing the proposed method, the program has been written in C++ language under Linux platform. The method was tested against DNA sequences of various sizes taken from NCBI databank(Table 5). The method provides the solution to problem of locating the exact position of the pattern in the text.

Table 5.Location of Pattern in Text

| Kennedy | (16,17,18);(53,54,55); |
|----------------------|--|
| disease | (59,60,61);(71,72,73); |
| | (101,102,103);(111,112,113); |
| Gene: AR | (136,137,138);(142,143,144); |
| | (153,154,155);(159,160,161); |
| Repeat motif: | (175,176,177);(363,364,365); |
| CAG | (372,373,374);(386,387,388); |
| , T | (441,442,443);(450,451,452); |
| Locus: | (463,464,465);(533,534,535); |
| NM_000044 | (554,555,556);(562,563,564); |
| Range: 1-700 | (663,664,665);(685,686,687); |
| Kalige. 1-700 | (689,690,691);(697,698,699) |
| | |
| Huntington | (33,34,35);(57,58,59); |
| disease | (196,197,198);(199,200,201); |
| | (202,203,204);(205,206,207); |
| | (208,209,210);(211,212,213); |
| 1 | |
| | (214,215,216);(217,218,219); |
| Gene: HD | (214,215,216);(217,218,219); (220,221,222);(223,224,225); |
| | |
| Repeat motif: | (220,221,222);(223,224,225); |
| | (220,221,222);(223,224,225); (226,227,228);(229,230,231); |
| Repeat motif: CAG | (220,221,222);(223,224,225); (226,227,228);(229,230,231); (232,233,234);(235,236,237); |
| Repeat motif: | (220,221,222);(223,224,225); (226,227,228);(229,230,231); (232,233,234);(235,236,237); (238,239,240);(241,242,243); |

| NM_002111 | (298,299,300);(307,308,309); |
|---------------|--|
| Range: 1-510 | (319,320,321);(325,326,327); (340,341,342);(346,347,348); |
| 8 | (428,429,430);(487,488,489); |
| | (495,496,497);(506,507,508) |
| | |
| Friedreich | (86,87,88);(151,152,153); |
| ataxia | (158,159,160);(232,233,234); |
| | (383,384,385);(391,392,393); |
| | (491,492,493);(508,509,510); |
| | (549,550,551);(816,817,818); |
| Gene: FRDA | (947,948,949);(1011,1012,1013 |
| Repeat motif: |); (1015,1016,1017);(1065,1066,1 |
| | 067); |
| GAA | (1300,1301,1302);(1304,1305,1 |
| т | 306); |
| Locus: | (1318,1319,1320);(1327,1328,1 |
| AH003505S1 | 329); |
| 71100550551 | (1345,1346,1347);(1378,1379,1 |
| Range: | 380); |
| U | (1485,1486,1487);(1535,1536,1 |
| 1-2465 | 537); |
| | (1558,1559,1560);(1601,1602,1 |
| | 603); |
| | (1631,1632,1633);(1639,1640,1 |
| | 641); |
| | (1649,1650,1651);(1668,1669,1 |
| | 670); (1711, 1712, 1712);(1722, 1724, 1 |
| | (1711,1712,1713);(1723,1724,1 |
| | 725); (1774,1775,1776);(1781,1782,1 |
| | 783); |
| | (1875,1876,1877);(1976,1977,1 |
| | 978); |
| | (1979,1980,1981);(2022,2023,2 |
| | 024); |
| | (2088,2089,2090);(2140,2141,2 |
| | 142); |
| | (2184,2185,2186);(2187,2188,2 |
| | 189); |
| | (2190,2191,2192);(2193,2194,2 |
| | 195); |
| | (2196,2197,2198);(2199,2200,2 |
| | 201); (2202 2203 2204)·(2205 2206 2 |
| | (2202,2203,2204);(2205,2206,2 207); |
| | (2208,2209,2210);(2217,2218,2 |
| | 219); |
| | (2282,2283,2284);(2427,2428,2 |
| | 429) |
| | |

We present a new sequential data mining method using correlation matrix memory. Here, we use the concept of Logical Match to locate the pattern in the text. This method can possibly be implement to develop a new approach related to the sequential data mining.

ACKNOWLEDGEMENT

SSKP was funded in part by European Research and Educational Collaboration with Asia

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V CONCLUSION