# Signal Processing on Databases

Jeremy Kepner

**Lecture 4: Analysis of Structured Data** 



This work is sponsored by the Department of the Air Force under Air Force Contract #FA8721-05-C-0002. Opinions, interpretations, recommendations and conclusions are those of the authors and are not necessarily endorsed by the United States Government.



#### **Outline**



- Introduction
  - Schema
  - Stats (Analytic 1)
- First Order Analytics
- Second Order Analytics
- Summary



### **Generic D4M Triple Store Schema**

#### **Accumulo Table: Ttranspose**

#### **Input Data**

Time	Col1	Col2	Col3
2001-01-01	а		а
2001-01-02	b	b	
2001-01-03		С	С

	01-01- 2001	02-01- 2001	03-01- 2001
Col1 a	1		
Col1 b		1	
Col2 b		1	
Col2 c			1
Col3 a	1		
Col3 c			1

	Col1 a	Col1 b	Col2 b	Col2 c	Col3 a	Col3 c
01-01-2001	1				1	
02-01-2001		1	1			
02 04 2004				4		4

#### **Accumulo Table: T**

- Tabular data expanded to create many type/value columns
- Transpose pairs allows quick look up of either row or column
- Big endian time for parallel performance



# Stats (Analytic 1) Diagram

Accum	nulo Table: T	ලි	(%)	Ç ON	<b>&gt;</b>	ශ්	الى ما	is cold	پر من	Ď	S	૾ૢૺ૾ૺૢૺ	ું જ	% '	93/9	တ်	Al <sup>®</sup> Cc	ialo c	olal <sup>c</sup> ol	xid cold	No.
Row	Key				Ţ					T					Ī						
1	01-10-2001 01 01 00				ļ					1											• • • • • •
2	01-10-2001 01 02 00																				••••
3	01-10-2001 01 03 00			A	<b>5</b> 50	<b>OC</b>	at	ve	Α	rra	ıy:	A									••••
4	01-10-2001 01 04 00									ļ											••••
5	01-10-2001 01 05 00																				
6	01-10-2001 01 06 00																				

- Copy a set of rows from T into associative array A
- Perform the following statistical calculations on A
  - Column count: how many times each column appears in A
  - Column type count: how many times each column type appears in A
  - Column covariance: how many times each pair of columns in A appear in the same row together
  - Column covariance: how many times each pair of column types in A appear in the same row together
- Good for identifying column types, gaps, clutter, and correlations

### **Stats Implementation**

Define a set of rows

 $r = '01-01-2001\ 01\ 02\ 00,01-01-2001\ 01\ 03\ 00,01-01-2001\ 01\ 04\ 00,$ 

Copy rows from table to associative array and convert '1' to 1

$$A = dblLogi(T(r,:))$$

Compute column counts

Compute column covariance

$$A' * A$$
 or  $sqln(A)$ 

Compute column type counts and covariance by substituting

$$A = col2type(A,'|');$$



#### **Outline**

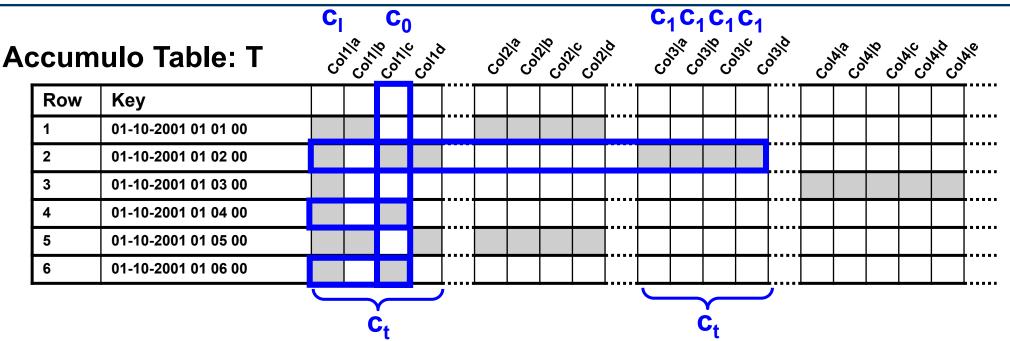
• Introduction



- First Order Analytics
  - Data Graph (Analytic 2)
  - Space (Analytic 3)
  - Convolution (Analytic 4)
- Second Order Analytics
- Summary



## Data Graphs (Analytic 2) Diagram



- Define data graph inputs
  - Start columns c<sub>0</sub>
  - Allowed column types C<sub>t</sub>
  - Clutter columns c
- Get all columns c<sub>1</sub> in rows containing c<sub>0</sub> of type c<sub>t</sub> and excluding columns c<sub>1</sub>
  - The fundamental operation upon which all graphs are built
  - Perform recursively to grow graph from starting columns



### **Data Graph Implementation**

Define start columns, allowed column types and clutter

Copy all columns from rows containing c0 into associative array

$$A = dblLogi(T(Row(T(:,c0)),:))$$

Reduce to allowed columns

$$A = A(:,ct)$$

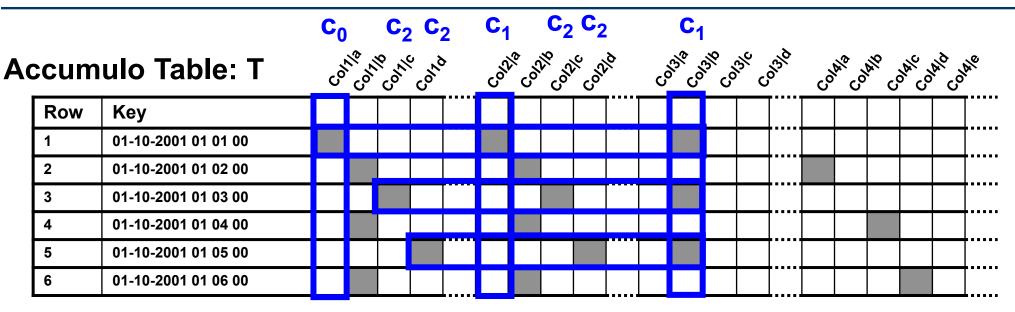
Eliminate clutter columns and return column labels

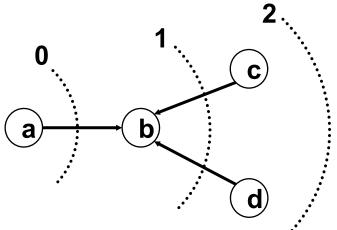
$$c1 = Col(A - A(:,cl))$$

Look for new clutter



### **Data Graphs Example 1**

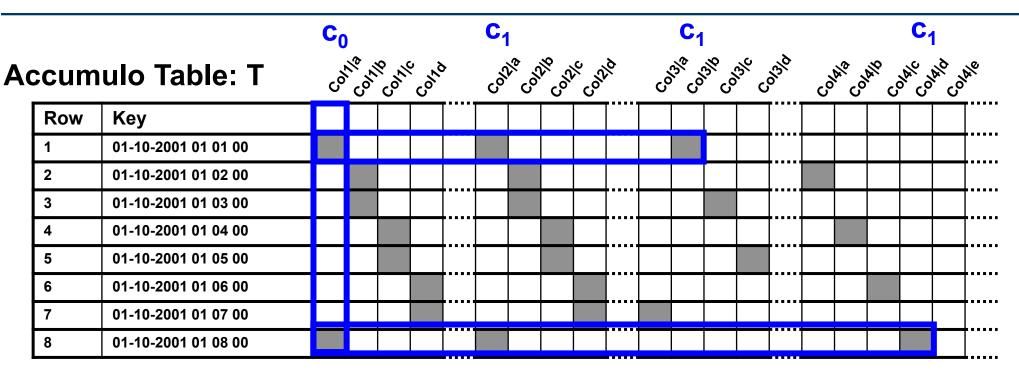


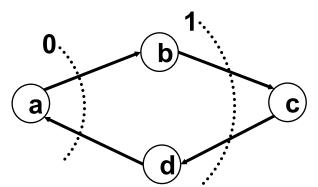


- Limited by the natural topology of the data
- Star data is good for generating star data graphs



### **Data Graphs Example 2**





- Limited by the natural topology of the data
- Star data is limiting for generating cycle data graphs



# Space (Analytic 3) Diagram

A	ccum	ulo Table: T	<b>C</b>		C %	C V V V	>	ું ડું	ان کا	ڒ؈ٛ	ું <sub>જો</sub>	٥,	40,	\$ 0,40 7,00	_	, 40°	<b>%</b>	70	S	S N	2 <sup>7</sup> / <sub>0</sub>	10°	ò
	Row	Key (time)					Ī																
	1	01-10-2001 01 01 00					<u> </u>																
	2	01-10-2001 01 02 00																					••••
r	3	01-10-2001 01 03 00	A	SS	OC	iat	ive	A	rra	ay:	Α												
	4	01-10-2001 01 04 00																					••••
	5	01-10-2001 01 05 00																					
	6	01-10-2001 01 06 00																					

- Select row range r and a space polygon s
- Copy a set of rows from T into associative array A
- Extract space coordinates from rows and determine if inside s
- Return columns c that satisfy these constraints
- Good for finding columns in a particular space window
- Can apply filter to space first is coordinates are "Mertonized"



### **Space Implementation**

Define row range and space polygon

```
r='01-01-2001 00 02 00,:,01-01-2001 00 04 00,'
s=complex([11 15 15 11 11],[15 15 11 11 15])
```

Copy all rows within t into associative array

$$A = T(r,:)$$

Get coordinates

$$Axy = str2num(col2type(A(:,StartsWith('X|,Y|,')),'|'))$$

Select columns in rows in space polygon



# Convolution (Analytic 4) Diagram

														C	C	C							
A	ccum	ulo Table: T	රි		(O)/	ري. ري.	<b>&gt;</b>	රු	المي محا	, co?	ير دمر	8	46	to to	, to	, to	_	,	24 0	bo t	30 48	1/2 1/2	
	Row	Key					Ţ					ļ											ļ
	1	01-10-2001 01 01 00																					<u> </u>
	2	01-10-2001 01 02 00																					
	3	01-10-2001 01 03 00	A	SS	OC	iat	ive	• 🖊	rra	ay:	Α	ļ											
	4	01-10-2001 01 04 00															••••						
	5	01-10-2001 01 05 00																					
	6	01-10-2001 01 06 00																					

- Copy a set of rows from T into associative array A
- Select a numeric column type and convolve with a filter

Standard signal processing technique for finding groups



### **Convolution Implementation**

Define a set of rows and a filter of width 4

```
r = 01-01-2001 01 02 00,01-01-2001 01 03 00,01-01-2001 01 04 00,

f = ones(1,4)
```

Copy rows from table to associative array and convert '1' to 1

$$A = dblLogi(T(r,:))$$

Create vector of numeric type rows

Convolve with filter and find columns > 1

$$c = Col(conv(Av,f) > 1)$$



#### **Outline**

- Introduction
- First Order Analytics



- Second Order Analytics
  - Type Pair (Analytic 5)
  - Data Pair (Analytic 6)
  - Semantic Extension (Analytic 7)
  - Semantic Pair (Analytic 8)
- Summary



# Type Pair (Analytic 5) Diagram

cumul			_											C					C		C	
	lo Table: T	လိ		(S)()		•	လိ	No O	ڒ؈؆	ير مي	<b>,</b>	40	o to	to,	* to	Ó	78	10 19	12 12	2, TO	10,00	8
Row K	<b>Cey</b>					Ţ										••••						<b>I</b> "
1 0 <sup>-</sup>	1-10-2001 01 01 00																					ľ
2 0	1-10-2001 01 02 00					ļ										••••						•
3 0′	1-10-2001 01 03 00	A	SS	OC	iat	ive	<b>}</b>	\rra	ay:	Α						••••						ľ
4 0	1-10-2001 01 04 00					Ť										••••						ľ
5 0′	1-10-2001 01 05 00					ļ										• • • • •						ľ
6 0 <sup>-</sup>	1-10-2001 01 06 00					1															П	1

- Copy a set of rows from T into associative array A
- Find rows in A that contain both pair types c<sub>t1</sub> and c<sub>t2</sub>
- Find columns of each type are paired with more than one column of the other type
- Good for tracking columns that occur in pairs

### **Type Pair Implementation**

Define row range and type pair

- Copy rows from table to associative array and convert '1' to 1
   A = dblLogi(T(r,:))
- Find rows containing both column types in the pair

$$r = Row(sum(A(Row(sum(A(:,ct1),2)==1),[ct1 ct2]),2)==2);$$

Get columns in order for creating a pair mapping matrix

$$[tmp c1 tmp] = A(r,ct1)$$
  
 $[tmp c2 tmp] = A(r,ct2)$   
 $A12 = Assoc(c1,c2,1)$ 

Find ct1 with more than one ct2 and vice versa

$$sum(A12,1) > 1$$
  $sum(A12,2) > 1$ 



# Data Pair (Analytic 6) Diagram

				C.	C.	1 C <sub>1</sub>							$\mathbf{C_2}$	$\mathbf{C_2}$	C <sub>2</sub>							
A	ccum	ulo Table: T	ď	51 <sup>1</sup> 201	(S)()	د کاره	<b>&gt;</b>	colli	COLL	Sollic (	<i>Solald</i>	ď	93% CO	<sup>130</sup> Co.	ي ن ک	13/0	رما	No Coli	نی	JAIC OLA	COLA	<b>,</b>
	Row	Key (time)									<b>—</b>											••••
	1	01-10-2001 01 01 00					••••															••••
r <sub>12</sub>	2	01-10-2001 01 02 00																				
	3	01-10-2001 01 03 00																				
	4	01-10-2001 01 04 00																				
	5	01-10-2001 01 05 00																				
	6	01-10-2001 01 06 00																				

- Define column pair sets c<sub>1</sub> and c<sub>2</sub>
- Get all columns c<sub>1</sub> and c<sub>2</sub>
- Find rows r<sub>12</sub> that have one entry in c<sub>1</sub> and c<sub>2</sub>

Checks to see if data pairs are present in the same row

### **Data Pair Implementation**

#### Define column pair sets

#### Create pair mapping matrices

$$A1p = Assoc(c1,c12,1) \qquad A2p = Assoc(c2,c12,1)$$

#### Get columns from T

$$A1 = dblLogi(T(:,c1))$$
  
 $A2 = dblLogi(T(:,c2))$ 

#### Find pairs

$$((A1*A1p) + (A2*A2p)) > 1)$$



# **Semantic Extension (Analytic 7)**

 Column types may have several types of semantic relationships which can be used to extend pairs

Pair reversal

Example: pair 'Col1|a;Col3|b' implies'Col3|b;Col1|a'

Type extension.

Example: column 'Col1|a' implies'Col2|a'

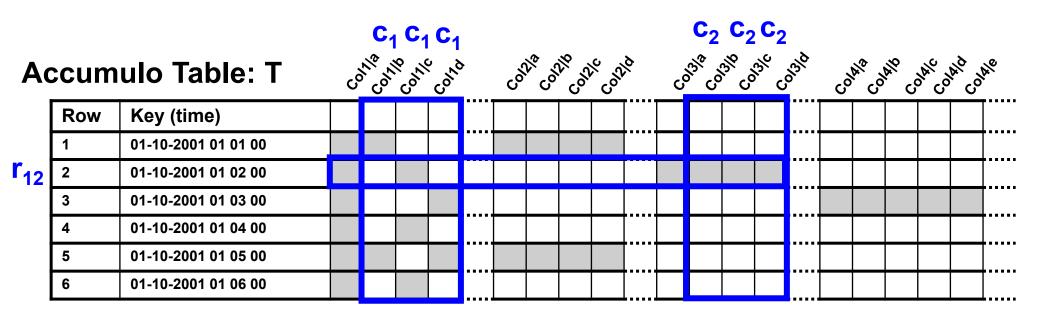
Data graph extension.

Example: column 'Col1|a' implies'Col2|b'if 'Col1|a' and Col2|b'appear in the same row

 Allows additional semantic data to be used to greatly increase the number columns that can be matched in a table



# Semantic Pair (Analytic 8) Diagram



- Define column pair sets c<sub>1</sub> and c<sub>2</sub>
- Extend all columns via semantic information
- Get all columns c<sub>1</sub> and c<sub>2</sub>
- Find rows r<sub>12</sub> that have one entry in c<sub>1</sub> and c<sub>2</sub>
- Checks to see if semantic pairs are present in the same row



### Summary

Exploded Schema allows rapid access to both rows and column

 Graph analytics can be implemented as a sequence of row and column queries

Complex analytics can be implemented via matrix multiply



### **Example Code & Assignment**

- Example Code (end of Lecture 3 and start of lecture 4)
  - d4m\_api/examples/2Apps/1EntityAnalysis
  - d4m\_api/examples/2Apps/2TrackAnalysis

- Assignment 3
  - For your associative arrays in Assignment 1 compute three different cross correlations using matrix multiply
  - Explain the meaning of each cross-correlation

MIT OpenCourseWare https://ocw.mit.edu

RES.LL-005 Mathematics of Big Data and Machine Learning IAP 2020

For information about citing these materials or our Terms of Use, visit: https://ocw.mit.edu/terms.