

CineCubes: Cubes as Movie Stars with Little Effort

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Univ. of Ioannina

Can we answer user queries with **more**
than just a set of records?

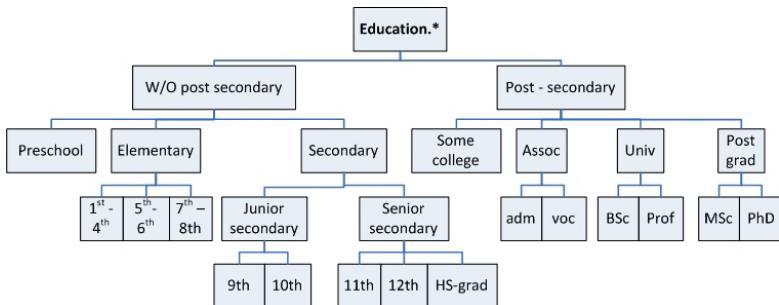
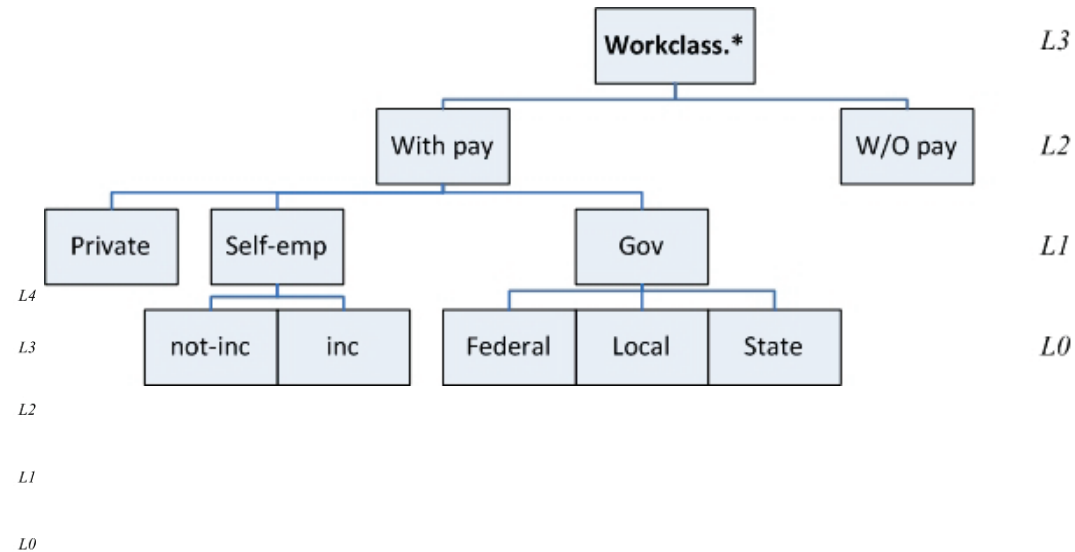
*We should and can produce query results that
are*

- *properly visualized,*
- *enriched with textual comments,*
- *vocally enriched,*

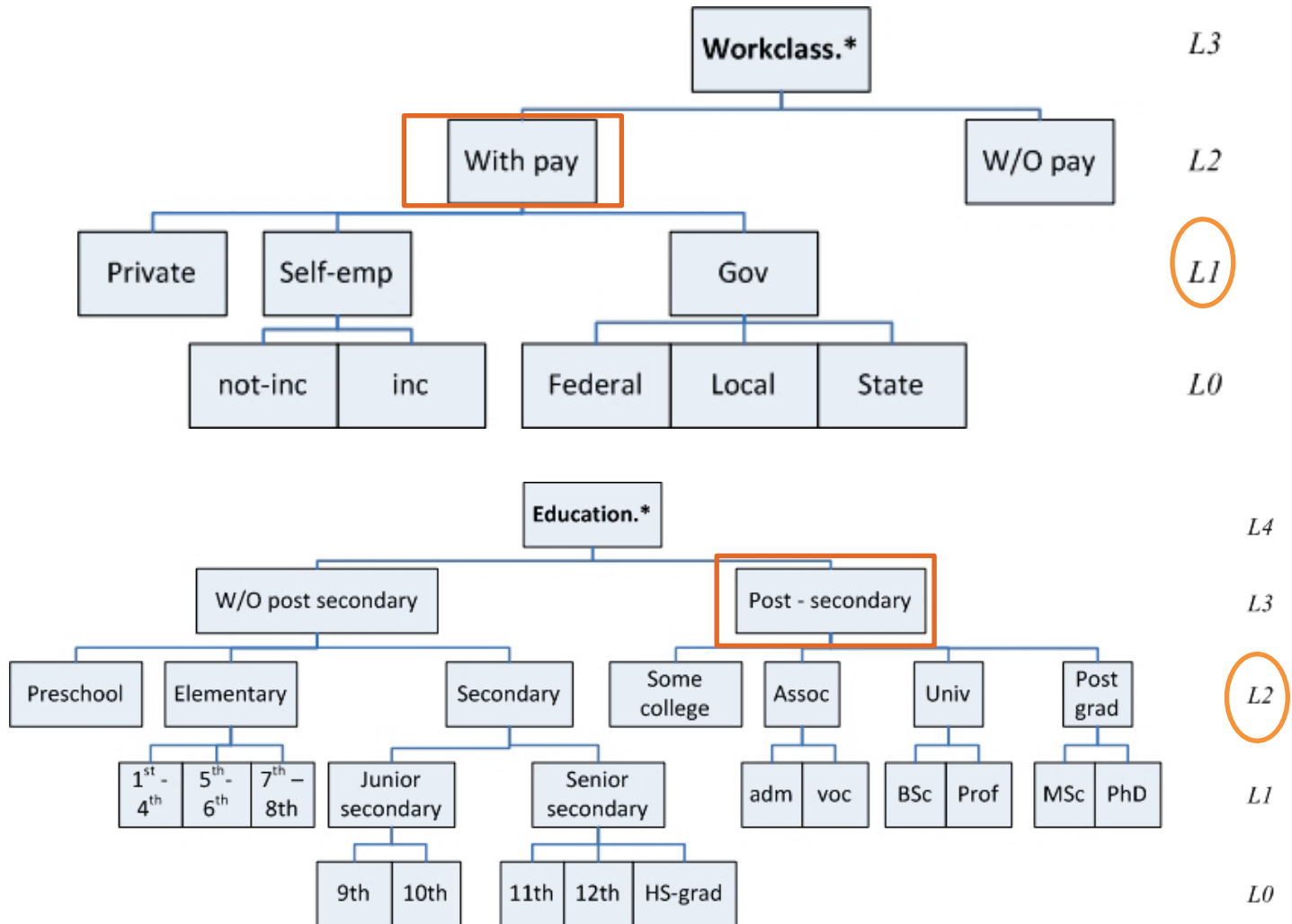
... but then, you have a movie

Example

- Find the average work hours per week *//measure*
 - For persons with *//selection conditions*
 - work_class.level2='With-Pay' , and
 - education.level3= 'Post-Sec'
 - Grouped per *//groupers*
 - work_class.level1
 - education.level3

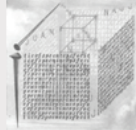


Example



Example: Result

CineCube Report



This is a report on the Age of the when for education at being to Post-secondary and work at being to Work-For. We will start by answering the original query and we complement the result with visualization and detailed analysis.

Answers the original question

	total	Normal	Knowledge	Unlikely
Age	0.75	0.68	0.68	0.53
Work	0.58	0.58	0.75	0.58
Unlikely	0.58	0.51	0.75	0.53

Act I: Putting results in context

In this series of slides we put the original result in context, by comparing the behavior of its defining values with the behavior of values that are similar to them.

Assessing the behavior of education

Learning to	Normal	Unlikely
Age	0.53	0.67
Work	0.58	0.53
Unlikely	0.58	0.51

Assessing the behavior of work

Learning to	total	Normal	Knowledge	Unlikely
Age	0.75	0.68	0.68	0.53
Work	0.58	0.58	0.75	0.58
Unlikely	0.58	0.51	0.75	0.53

Act II: Explaining results

In this series of slides we will present a detailed analysis of the values involved in the results of the original query. To this end, we will attempt to drill-down the hierarchy of grouping levels of the result to one level of aggregation lower whenever is possible.

Answers the original question

	total	Normal	Knowledge	Unlikely
Age	0.75	0.68	0.68	0.53
Work	0.58	0.58	0.75	0.58
Unlikely	0.58	0.51	0.75	0.53

Drilling down the Rows of the Original Result

Age	total	Normal	Knowledge	Unlikely
Post-high	0.53 (68)	0.68 (81)	0.68 (81)	0.53 (68)
College	0.53 (75)	0.68 (85)	0.68 (85)	0.53 (85)
Somecol	0.53 (47)	0.68 (58)	0.68 (58)	0.53 (47)

Work	total	Normal	Knowledge	Unlikely
Home	0.58 (75)	0.58 (88)	0.75 (93)	0.58 (88)
Unlikely	0.58 (23)	0.58 (23)	0.75 (23)	0.58 (23)

Unlikely	total	Normal	Knowledge	Unlikely
Self-employed	0.58 (75)	0.58 (88)	0.75 (93)	0.58 (88)
Self-employed	0.58 (23)	0.58 (23)	0.75 (23)	0.58 (23)

Drilling down the Columns of the Original Result

Age	total	Normal	Knowledge	Unlikely
Post-high	0.53 (68)	0.68 (81)	0.68 (81)	0.53 (68)
College	0.53 (75)	0.68 (85)	0.68 (85)	0.53 (85)
Somecol	0.53 (47)	0.68 (58)	0.68 (58)	0.53 (47)

Work	total	Normal	Knowledge	Unlikely
Home	0.58 (75)	0.58 (88)	0.75 (93)	0.58 (88)
Unlikely	0.58 (23)	0.58 (23)	0.75 (23)	0.58 (23)

Unlikely	total	Normal	Knowledge	Unlikely
Self-employed	0.58 (75)	0.58 (88)	0.75 (93)	0.58 (88)
Self-employed	0.58 (23)	0.58 (23)	0.75 (23)	0.58 (23)

Summary

- Concerning the original query we can conclude the following results:
 - Column Knowledge has the highest values.
 - Row Self-employed has the highest values.
 - Row Self-employed has the highest values.
- When we drill-down and the original result for work, by comparing its defining values with similar ones, we discovered the following:
 - Expanded to its defining values in the end of the column Knowledge has the highest values than the other values.
 - Expanded to its defining values in the end of the column Knowledge has the highest values than the other values.
- In fact, it is seen that the row Self-employed has the highest values.
- In fact, it is seen that the row Self-employed has the highest values.
- When we drill-down the results of the original query, we discovered the following:
 - Column Knowledge has the highest values.
 - Column Self-employed has the highest values.



Answer to the original question

	Assoc	Post-grad	Some- college	University
Gov	40.73	43.58	38.38	42.14
Private	41.06	45.19	38.73	43.06
Self-emp	46.68	47.24	45.70	46.61

Here, you can see the answer of the original query. You have specified education to be equal to 'Post-Secondary', and work to be equal to 'With-Pay'. We report on Avg of work hours per week grouped by education at level 2, and work at level 1 .

You can observe the results in this table. We highlight the largest values with red and the lowest values with blue color.

Column Some-college has 2 of the 3 lowest values.

Row Self-emp has 3 of the 3 highest values.

Row Gov has 2 of the 3 lowest values.

1	Assoc	Post-grad	Some-college	University
Gov	40.73	43.58	38.38	42.14
Private	41.06	45.19	38.73	43.06
Self-emp	46.68	47.24	45.70	46.61

Original query

Here, you can see the answer of the original query. You have specified education to be equal to 'Post-Secondary', and work to be equal to 'With-Pay'. We report on Avg of Hrs grouped by education at level 2, and work at level 1. We highlight the largest values with red and the lowest values with blue.

Column Some-college has 2 of the 3 lowest values.
Row Self-emp has 3 of the 3 highest values.
Row Gov has 2 of the 3 lowest values.

Drilling down education

5	Assoc	Gov	Private	Self-emp
	Assoc-acdm	39.91 (182)	40.87 (720)	45.49 (105)
	Assoc-voc	41.61 (169)	41.20 (993)	47.55 (145)
	Post-grad	Gov	Private	Self-emp
	Doctorate	46.53 (124)	49.05 (172)	47.22 (79)
	Masters	42.93 (567)	44.42 (863)	47.25 (197)
	Some-college	Gov	Private	Self-emp
	Some-college	38.38 (955)	38.73 (5016)	45.70 (704)
	University	Gov	Private	Self-emp
	Bachelors	41.56 (943)	42.71 (3455)	46.23 (646)
	Prof-school	48.40 (86)	47.96 (247)	47.78 (209)

2	Post-Secondary	Without Post-Secondary
Gov	41.12	38.97
Private	41.06	39.40
Self-emp	46.39	44.84

Summary for education

Act I (sl. 2,3)

In this slide, we drill-down one level for all values of dimension work at level 0. For each cell we show both the Avg of Hrs and the number of tuples that correspond to it in parentheses. ...
Column Post-grad has 4 of the 6 highest values.
Column Some-college has 4 of the 6 lowest values.

In this graphic, we put the original request in context by comparing the value 'Post-Secondary' for education at level 3 with its sibling values. We calculate the Avg of Hrs while fixing education at level 4 to be equal to 'ALL', and work at level 2 to be equal to 'With-Pay'. We highlight the reference cells with bold, the highest value with red and the lowest value with blue.
Compared to its sibling we observe that in 3 out of 3 cases Post-Secondary has higher value than Without-Post-Secondary.

3	Assoc	Post-grad	Some-college	University
With-Pay	41.62	44.91	39.41	43.44
Without-pay	50.00	-	35.33	-

Summary for work

Drilling down work

4	Gov	Assoc	Post-grad	Some-college	University
	Federal-gov	41.15 (93)	43.86 (80)	40.31 (251)	43.38 (233)
	Local-gov	41.33 (171)	43.96 (362)	40.14 (385)	42.34 (499)
	State-gov	39.09 (87)	42.93 (249)	34.73 (319)	40.82 (297)
	Private	Assoc	Post-grad	Some-college	University
	Private	41.06 (1713)	45.19 (1035)	38.73 (5016)	43.06 (3702)
	Self-emp	Assoc	Post-grad	Some-college	University
	Self-emp-inc	48.68 (72)	53.05 (110)	49.31 (223)	49.91 (338)
	Self-emp-not-inc	45.88 (178)	43.39 (166)	44.03 (481)	44.44 (517)

Contributions

- We create a small “movie” that answers an OLAP query
- We complement each query with **auxiliary queries** organized in thematically related **acts** that allow us to assess and explain the results of the original query
- We implemented an extensible palette of **highlight** extraction methods to find interesting patterns in the result of each query
- We describe each highlight with **text**
- We use TTS technology to convert text to **audio**

Contributions

- Equally importantly:
 - An **extensible software** where algorithms for **query generation** and **highlight extraction** can be plugged in
 - The demonstration **of low technical barrier** to produce CineCube reports

Method Overview

Software Issues

Discussion

Method Overview

Our Approach

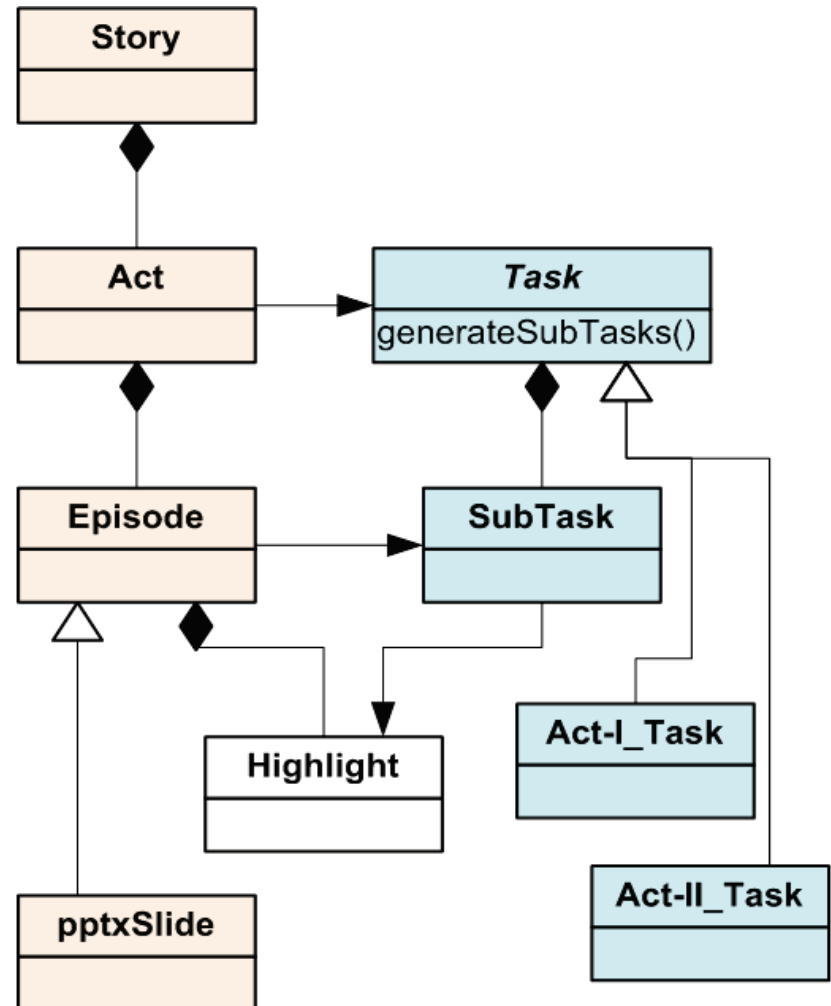
- A first assessment of the current state of affairs
 - Practically, this requirement refers to the execution of the original query.
- Put the state in Context
- Analysis of why things are this way.

Structure of the CineCube Movie

- A typical movie story is structured in **acts**.
 - Each Act is composed of sequences of scenes
- That's why we organize the CineCube Movie in five Acts:
 - Intro Act
 - Original Act
 - Act I
 - Act II
 - Summary Act

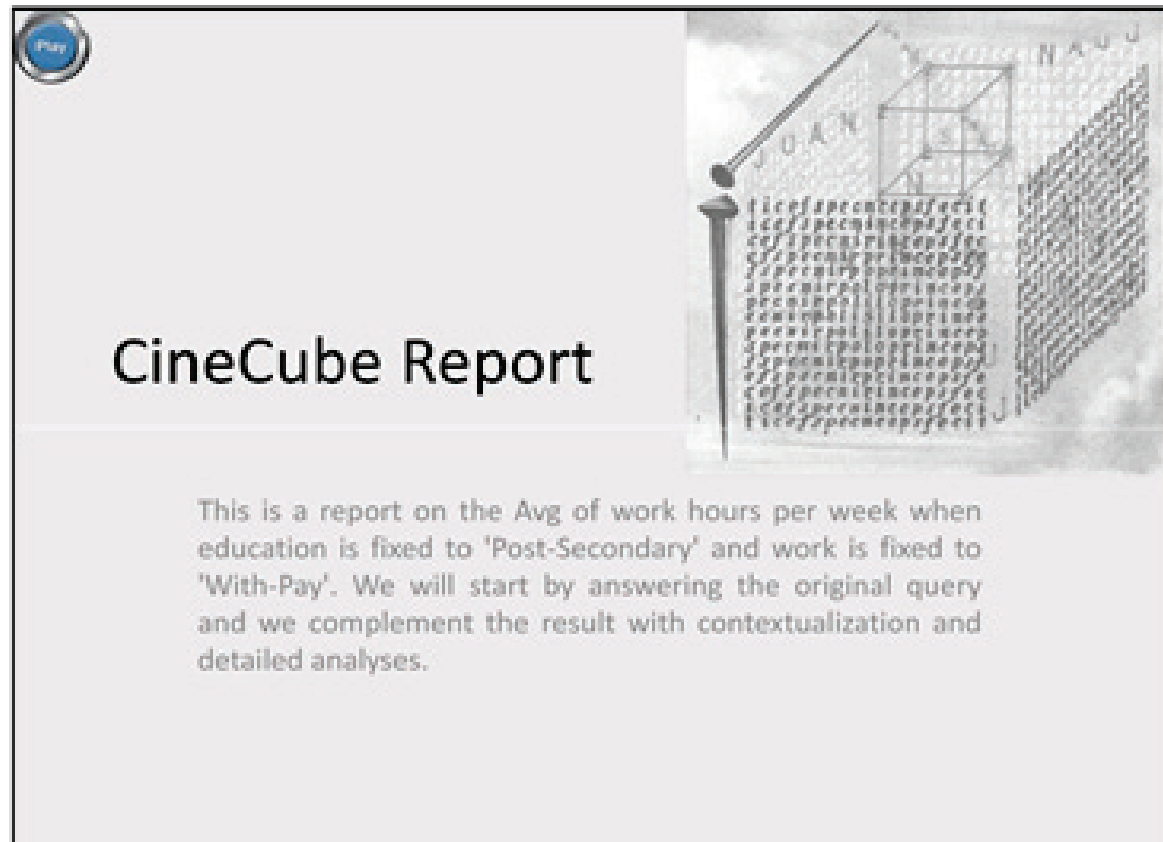
The movie's parts

- Much like movies, we organize our stories in acts
- Each act including several episodes all serving the same purpose



CineCube Movie – Intro Act

- Intro Act has an episode that introduce the story to user



CineCube Report

This is a report on the Avg of work hours per week when education is fixed to 'Post-Secondary' and work is fixed to 'With-Pay'. We will start by answering the original query and we complement the result with contextualization and detailed analyses.

CineCube Movie – Original Act

- Original Act has an episode which is the answer of query that submitted by user



The screenshot displays a table titled "Answer to the original question" within a software interface. The table has five columns: "Gov", "Assoc", "Post grad", "Some college", and "University". The rows represent employment types: "Gov", "Private", and "Self-emp". The numerical values in the table are color-coded: blue for "Gov" and "Some college", red for "Self-emp", and black for "Private".

	Assoc	Post grad	Some college	University
Gov	40.73	43.58	38.38	42.14
Private	41.06	45.19	38.73	43.06
Self-emp	46.68	47.24	45.70	46.61

CineCube Movie – Act I

- In this Act we try to answer the following question:
 - How good is the original query compared to its siblings?
- We compare the marginal aggregate results of the original query to the results of “sibling” queries that use “similar” values in their selection conditions

Act I – Example

Result of Original Query

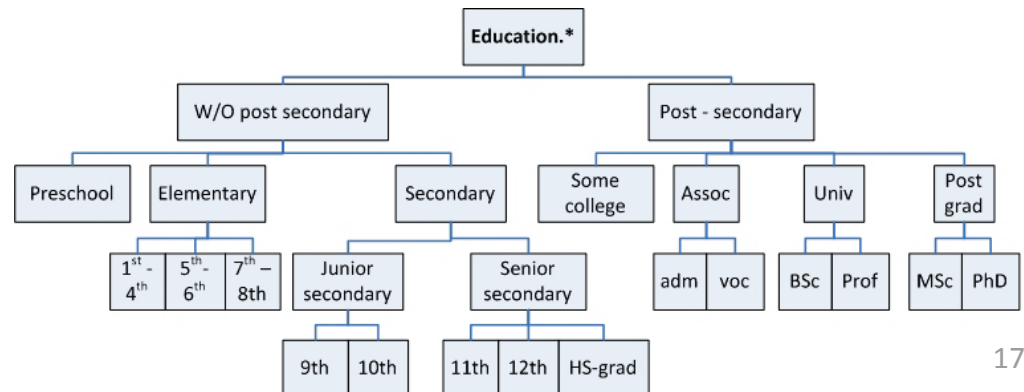
	Assoc	Post-grad	Some-college	University
Gov	40.73	43.58	38.38	42.14
Private	41.06	45.19	38.73	43.06
Self-emp	46.68	47.24	45.70	46.61

$q = (DS^0,$
 $W.L_2 = \text{'With-Pay'} \wedge E.L_3 = \text{'Post-Sec'},$
 $[W.L_1, E.L_2],$
 $\text{avg(Hrs)})$

Assessing the behavior of education

Summary for education	Post-Secondary	Without-Post-Secondary
Gov	41.12	38.97
Private	41.06	39.40
Self-emp	46.39	44.84

$q = (DS^0,$
 $W.L_2 = \text{'With-Pay'} \wedge E.L_4 = \text{'All'},$
 $[W.L_1, E.L_3],$
 $\text{avg(Hrs)})$



L4

L3

L2

L1

Act I – Example

Result of Original Query

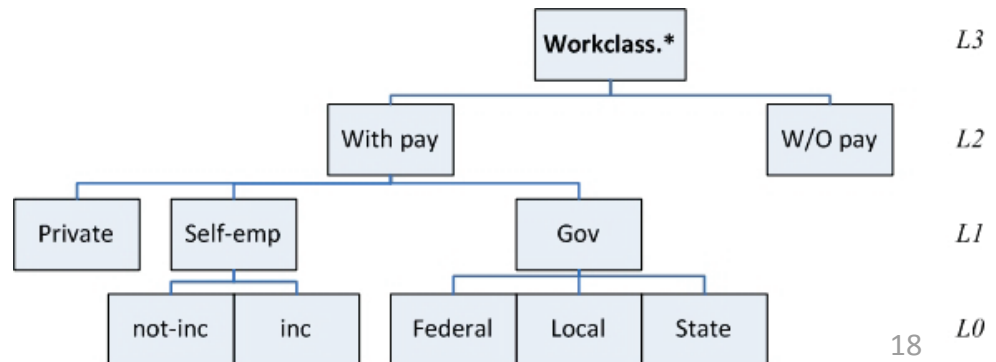
	Assoc	Post-grad	Some-college	University
Gov	40.73	43.58	38.38	42.14
Private	41.06	45.19	38.73	43.06
Self-emp	46.68	47.24	45.70	46.61

$q = (DS^0,$
 $W.L_2 = \text{'With-Pay'} \wedge E.L_3 = \text{'Post-Sec'},$
 $[W.L_1, E.L_2],$
 $\text{avg(Hrs)})$

Assessing the behavior of work

Summary for work	Assoc	Post-grad	Some-college	University
With-Pay	41.62	44.91	39.41	43.44
Without-pay	50.00	-	35.33	-

$q = (DS^0,$
 $W.L_3 = \text{'All'} \wedge E.L_3 = \text{'Post-Sec'},$
 $[W.L_2, E.L_2],$
 $\text{avg(Hrs)})$



CineCube Movie – Act II

- In this Act we try to explaining to user why the result of original query is what it is.
 - “Drilling into the breakdown of the original result”
- We drill in the details of the cells of the original result in order to inspect the internals of the aggregated measures of the original query.

Act II – Example

Result of Original Query

	Assoc	Post-grad	Some-college	University	
Gov	40.73	43.58	38.38	42.14	$q = (DS^0,$ $W.L_2 = \text{'With-Pay'} \wedge E.L_3 = \text{'Post-Sec'},$ $[W.L_1, E.L_2],$ $\text{avg(Hrs)})$
Private	41.06	45.19	38.73	43.06	
Self-emp	46.68	47.24	45.70	46.61	

Drilling down the Rows of the Original Result

		Assoc	Post-grad	Some-college	University
Gov	Federal-gov	41.15 (93)	43.86 (80)	40.31 (251)	43.38 (233)
	Local-gov	41.33 (171)	43.96 (362)	40.14 (385)	42.34 (499)
	State-gov	39.09 (87)	42.93 (249)	34.73 (319)	40.82 (297)
Private	Private	41.06 (1713)	45.19 (1035)	38.73 (5016)	43.06 (3702)
Self-emp	Self-emp-inc	48.68 (72)	53.05 (110)	49.31 (223)	49.91 (338)
	Self-emp-not-inc	45.88 (178)	43.39 (166)	44.03 (481)	44.44 (517)

Act II – Example

Result of Original Query

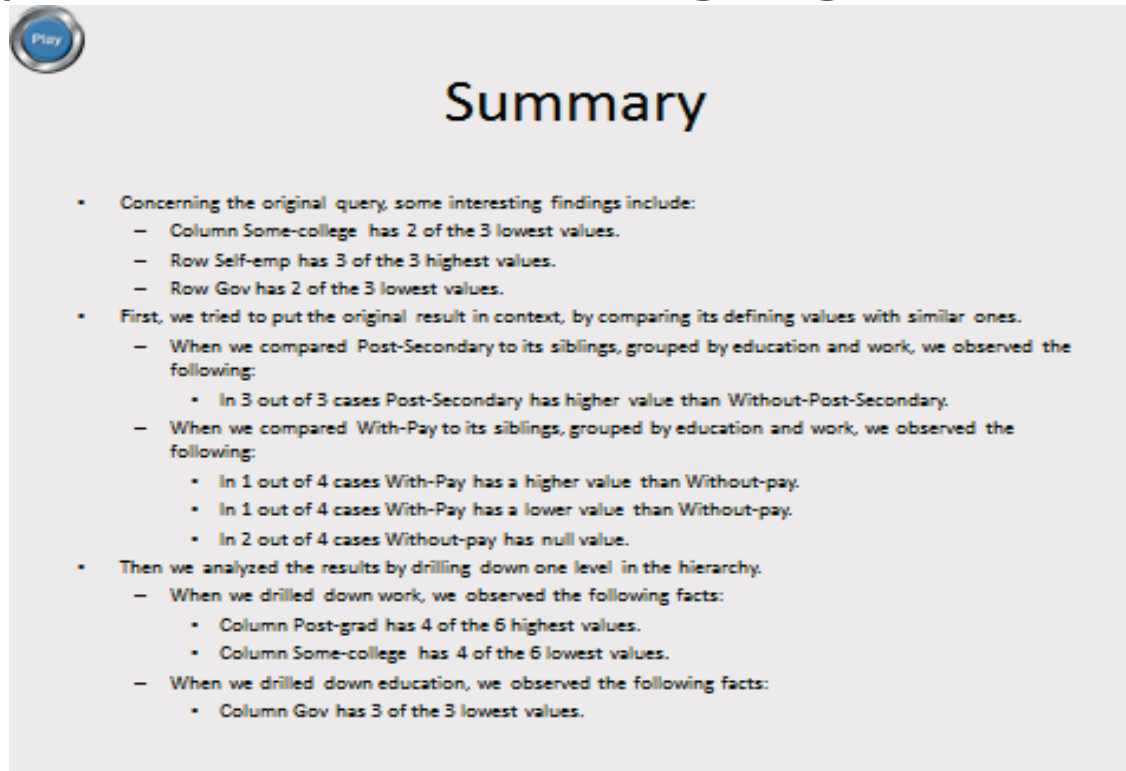
	Assoc	Post-grad	Some-college	University	
Gov	40.73	43.58	38.38	42.14	$q = (DS^0,$ $W.L_2 = \text{'With-Pay'} \wedge E.L_3 = \text{'Post-Sec'},$ $[W.L_1, E.L_2],$ $\text{avg(Hrs)})$
Private	41.06	45.19	38.73	43.06	
Self-emp	46.68	47.24	45.70	46.61	

Drilling down the Columns of the Original Result

	Assoc	Gov	Private	Self-emp
	Assoc-acdm	39.91 (182)	40.87 (720)	45.49 (105)
	Assoc-voc	41.61 (169)	41.20 (993)	47.55 (145)
Post-grad	Doctorate	46.53 (124)	49.05 (172)	47.22 (79)
	Masters	42.93 (567)	44.42 (863)	47.25 (197)
Some-college	Some-college	38.38 (955)	38.73 (5016)	45.70 (704)
University	Bachelors	41.56 (943)	42.71 (3455)	46.23 (646)
	Prof-school	48.40 (86)	47.96 (247)	47.78 (209)

CineCube Movie – Summary Act

- Summary Act represented from one episode.
- This episode has all the highlights of our story.



Summary

- Concerning the original query, some interesting findings include:
 - Column Some-college has 2 of the 3 lowest values.
 - Row Self-emp has 3 of the 3 highest values.
 - Row Gov has 2 of the 3 lowest values.
- First, we tried to put the original result in context, by comparing its defining values with similar ones.
 - When we compared Post-Secondary to its siblings, grouped by education and work, we observed the following:
 - In 3 out of 3 cases Post-Secondary has higher value than Without-Post-Secondary.
 - When we compared With-Pay to its siblings, grouped by education and work, we observed the following:
 - In 1 out of 4 cases With-Pay has a higher value than Without-pay.
 - In 1 out of 4 cases With-Pay has a lower value than Without-pay.
 - In 2 out of 4 cases Without-pay has null value.
- Then we analyzed the results by drilling down one level in the hierarchy.
 - When we drilled down work, we observed the following facts:
 - Column Post-grad has 4 of the 6 highest values.
 - Column Some-college has 4 of the 6 lowest values.
 - When we drilled down education, we observed the following facts:
 - Column Gov has 3 of the 3 lowest values.

Highlight Extraction

- We utilize a palette of highlight extraction methods that take a 2D matrix as input and produce important findings as output.
- Such as:
 - The top and bottom quartile of values in a matrix
 - The absence of values from a row or column
 - The domination of a quartile by a row or a column
 - The identification of min and max values

Text Extraction

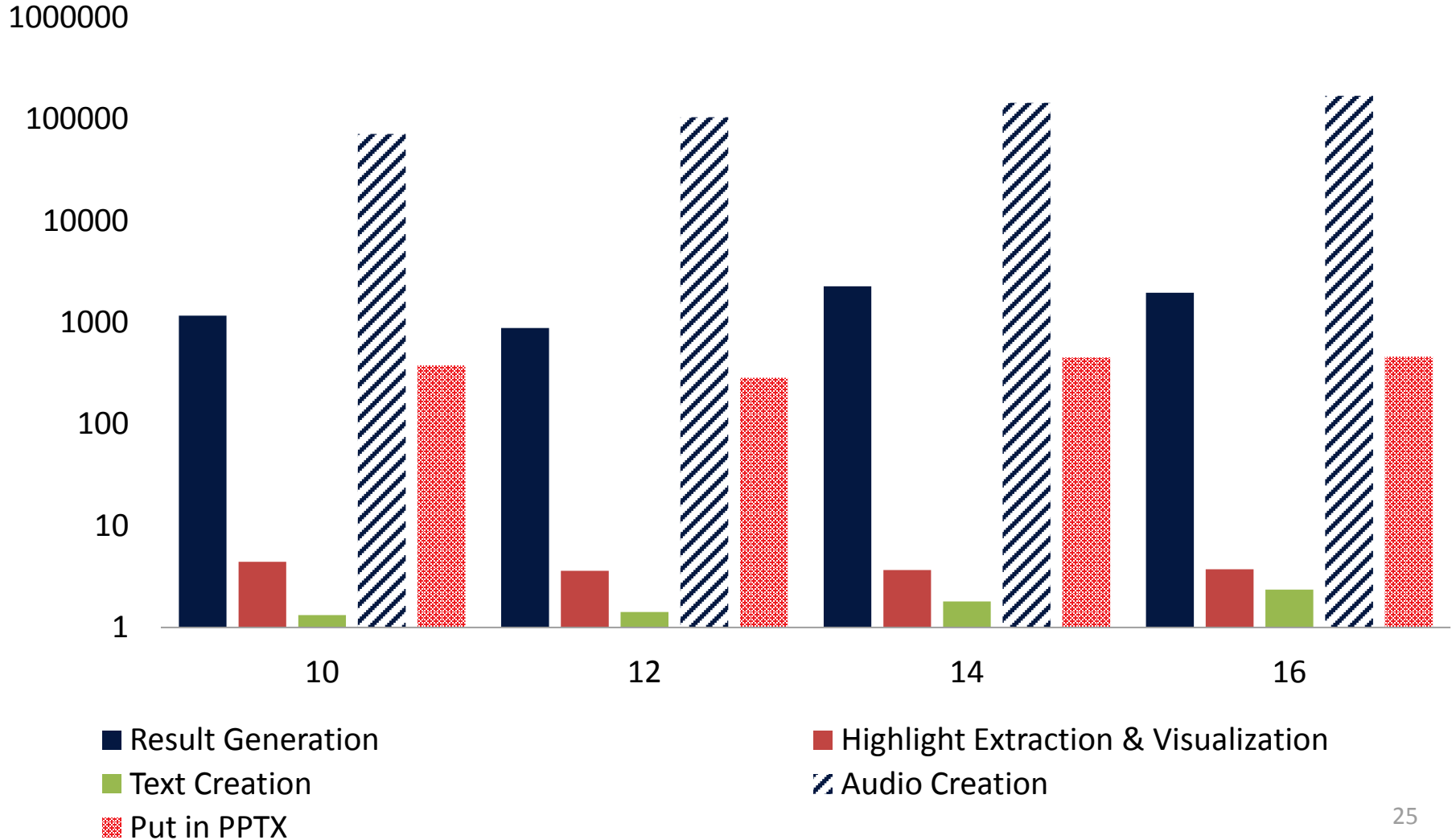
- Text is constructed by a Text Manager that customizes the text per Act

- **Example:**

In this slide, we drill-down one level for all values of dimension <dim> at level <l>. For each cell we show both the <agg> of <measure> and the number of tuples that correspond to it.

Experimental Results

Time breakdown(msec, log scale) for the method's parts

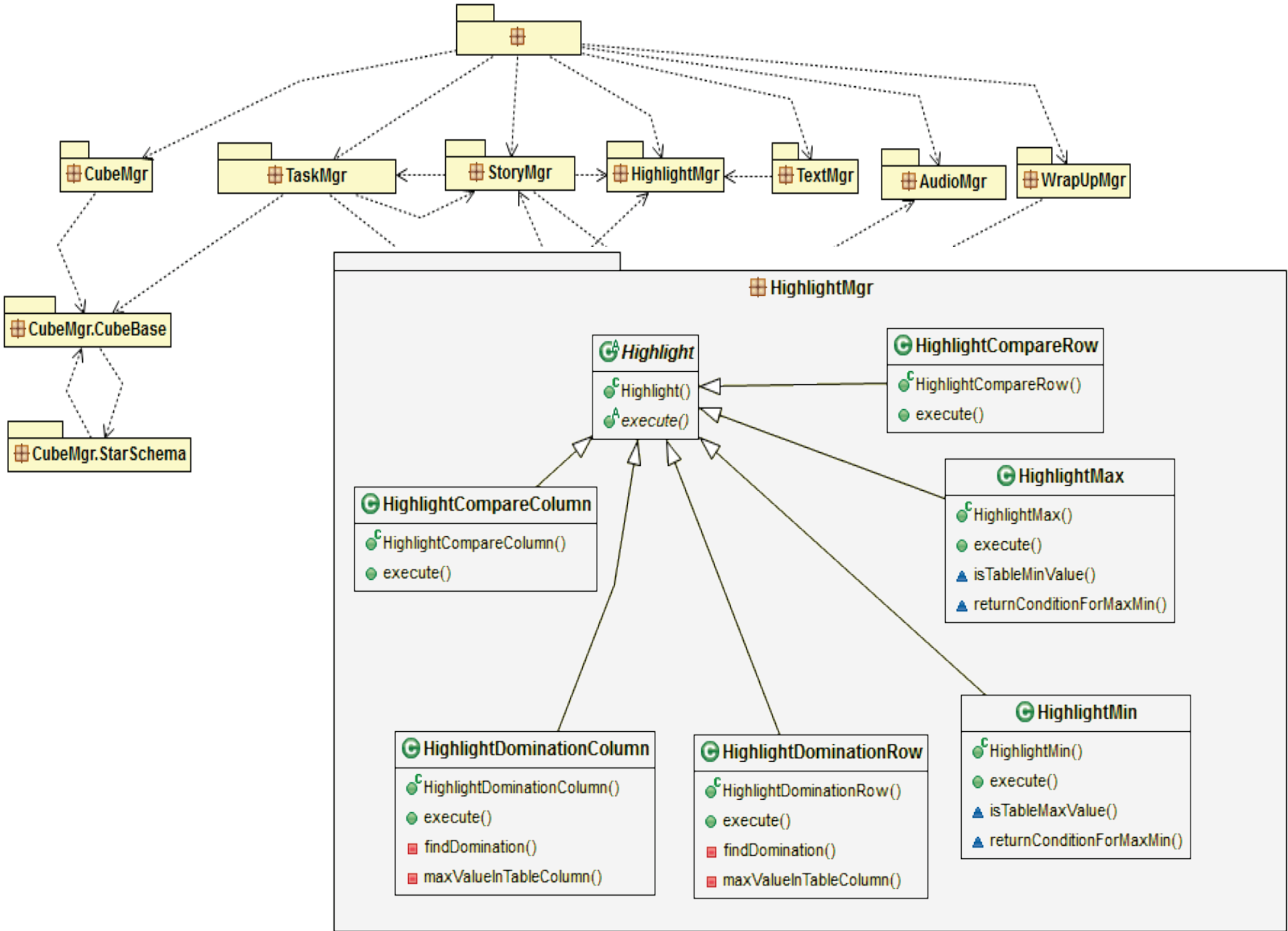


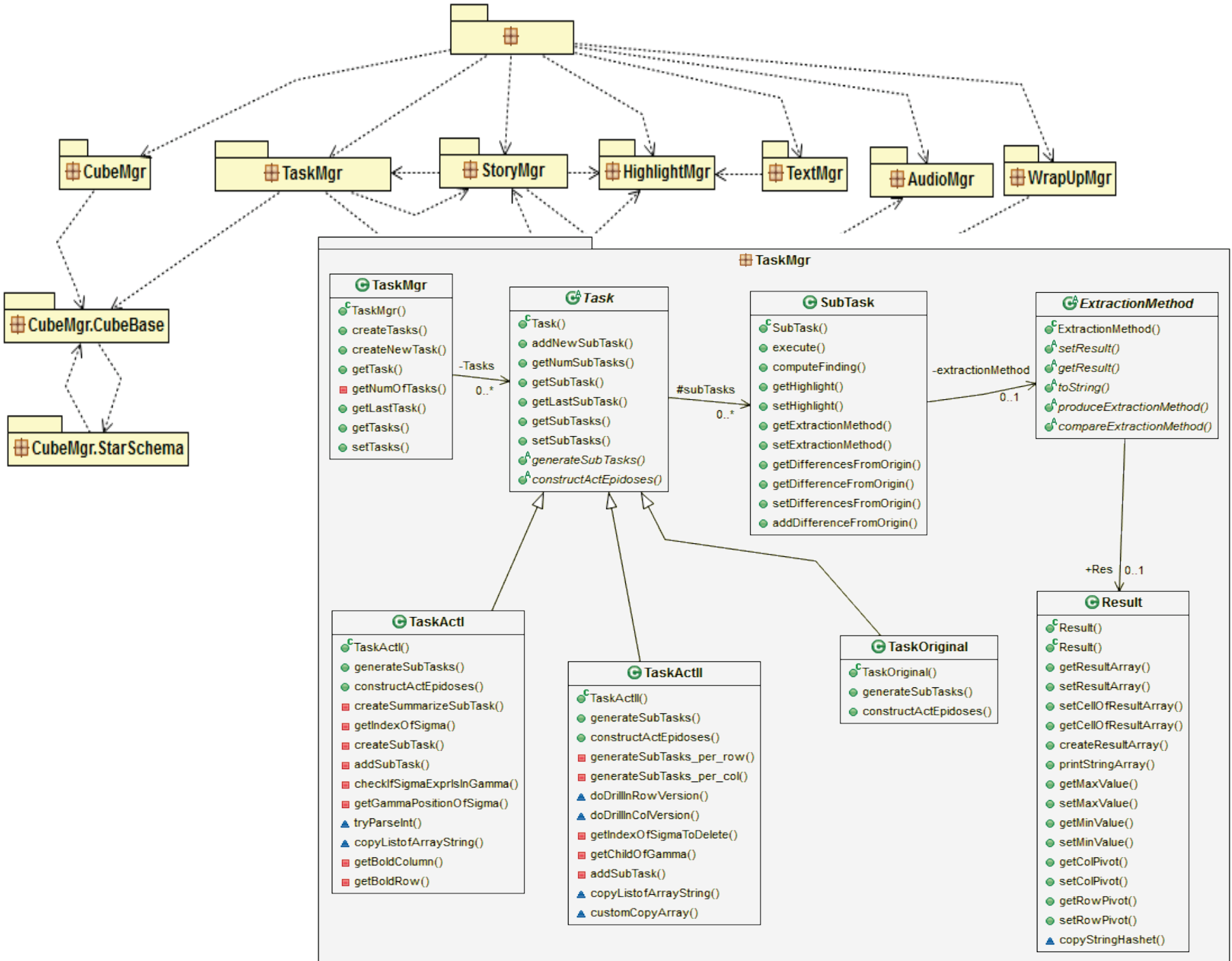
Method Overview
Software Issues
Discussion

Software Issues

Low technical barrier

- Our tool is extensible
 - We can add new tasks to generate complementary queries easily
 - We can add new highlight algorithms to produce highlights easily
- Supportive technologies are surprisingly easier to use
 - Apache POI for pptx generation
 - TTS for text to speech conversion





Apache POI for pptx

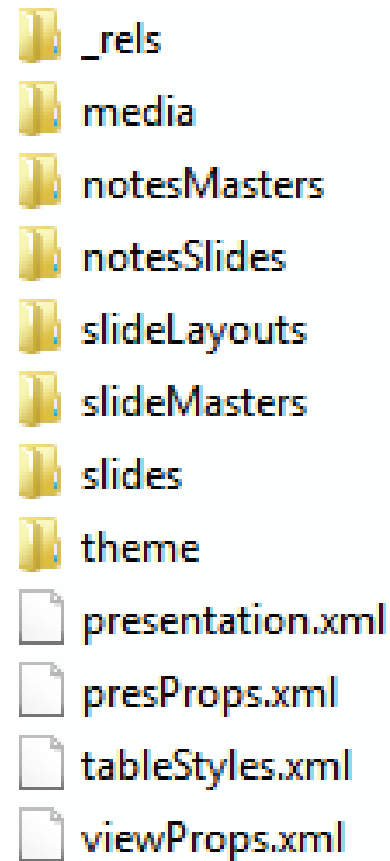
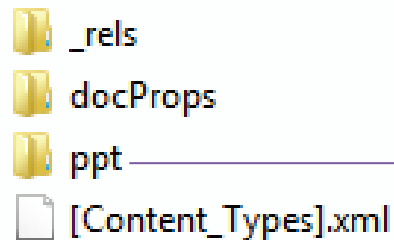
- A Java API that provides several libraries for Microsoft Word, PowerPoint and Excel (since 2001).
- XSLF is the Java implementation of the PowerPoint 2007 OOXML (.pptx) file format.

```
XMLSlideShow ss = new XMLSlideShow();  
XSLFSlideMaster sm = ss.getSlideMasters()[0];
```

```
XSLFSlide sl= ss.createSlide  
(sm.getLayout(SlideLayout.TITLE_AND_CONTENT));
```

```
XSLFTable t = sl.createTable();  
t.addRow().addCell().setText("added a cell");
```

PPTX Folder Structure



MaryTTS for Text-to-Speech Synthesis

```
MaryInterface m = new LocalMaryInterface();  
m.setVoice("cmu-slt-hsmm");  
  
AudioInputStream audio = m.generateAudio("Hello");  
  
AudioSystem.write(audio, audioFileFormat.Type.WAVE,  
new File("myWav.wav"));
```

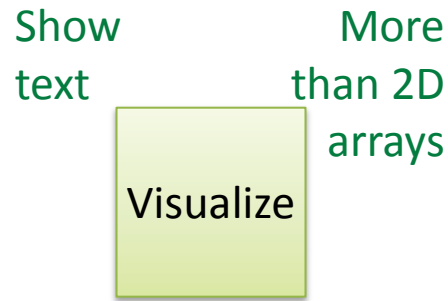
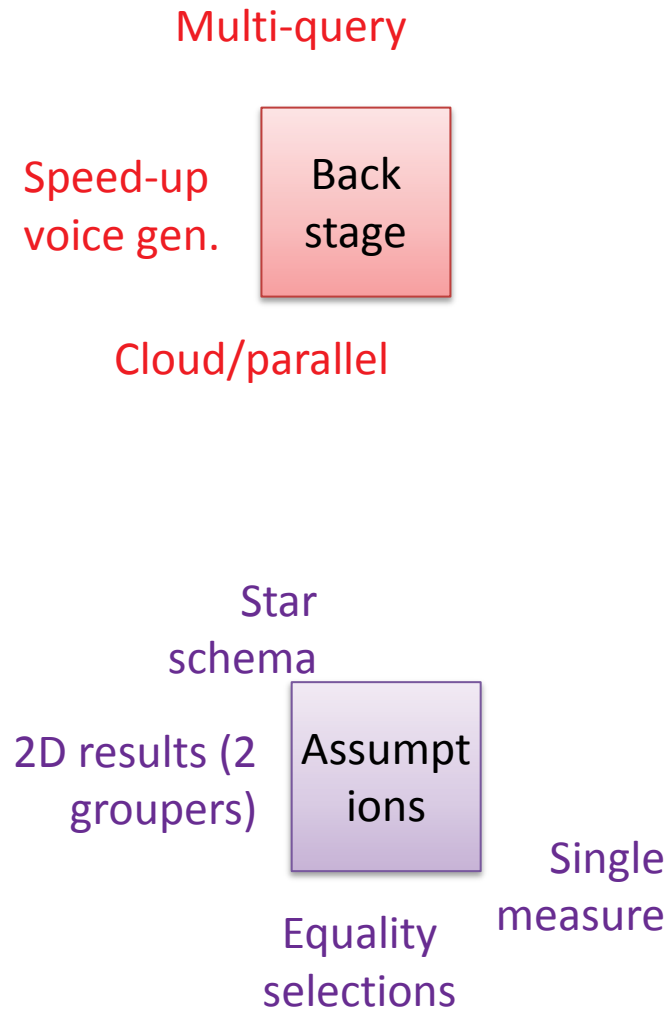

Method Overview

Software Issues

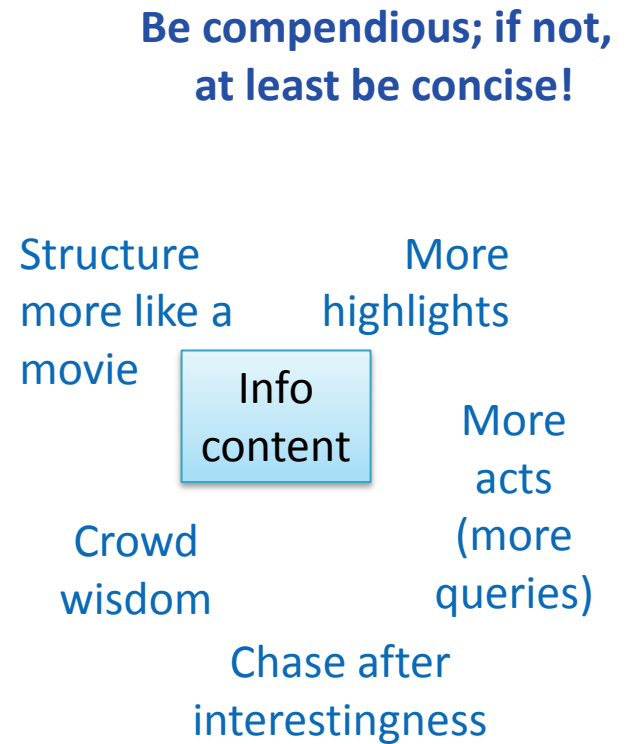
Discussion

Discussion

Open issues



Look like a movie



How to allow interaction with the user?

Personalization

Thank you!

Any questions?

More information

- <http://www.cs.uoi.gr/~pvassil/projects/cinecubes/>

Demo

- <http://snf-56304.vm.oceanos.grnet.gr/>

Code

- <https://github.com/DAINTINESS-Group/CinecubesPublic.git>

AUXILIARY SLIDES

Related Work

Related Work

- Query Recommendations
- Database-related efforts
- OLAP-related methods
- Advanced OLAP operators
- Text synthesis from query results

Query Recommendations

- A. Giacometti, P. Marcel, E. Negre, A. Soulet, 2011. Query Recommendations for OLAP Discovery-Driven Analysis. IJDWM 7,2 (2011), 1-25 DOI= <http://dx.doi.org/10.4018/jdwm.2011040101>
- C. S. Jensen, T. B. Pedersen, C. Thomsen, 2010. Multidimensional Databases and Data Warehousing. Synthesis Lectures on Data Management, Morgan & Claypool Publishers
- A. Maniatis, P. Vassiliadis, S. Skiadopoulos, Y. Vassiliou, G. Mavrogonatos, I. Michalarias, 2005. A presentation model and non-traditional visualization for OLAP. IJDWM, 1,1 (2005), 1-36. DOI= <http://dx.doi.org/10.4018/jdwm.2005010101>
- P. Marcel, E. Negre, 2011. A survey of query recommendation techniques for data warehouse exploration. EDA (Clermont-Ferrand, France, 2011), pp. 119-134

Database-related efforts

- K. Stefanidis, M. Drosou, E. Pitoura, 2009. "You May Also Like" Results in Relational Databases. PersDB (Lyon, France, 2009).
- G. Chatzopoulou, M. Eirinaki, S. Koshy, S. Mittal, N. Polyzotis, J. Varman, 2011. The QueRIE system for Personalized Query Recommendations. IEEE Data Eng. Bull. 34,2 (2011), pp. 55-60

OLAP-related methods

- V. Cariou, J. Cubillé, C. Derquenne, S. Goutier, F.Guisnel, H. Klajnmic, 2008. Built-In Indicators to Discover Interesting Drill Paths in a Cube. DaWaK (Turin, Italy, 2008), pp. 33-44, DOI=http://dx.doi.org/10.1007/978-3-540-85836-2_4
- A. Giacometti, P. Marcel, E. Negre, A. Soulet, 2011. Query Recommendations for OLAP Discovery-Driven Analysis. IJDWM 7,2 (2011), 1-25 DOI= <http://dx.doi.org/10.4018/jdwm.2011040101>

Advanced OLAP operators

- Sunita Sarawagi: User-Adaptive Exploration of Multidimensional Data. VLDB 2000:307-316
- S. Sarawagi, 1999. Explaining Differences in Multidimensional Aggregates. VLDB (Edinburgh, Scotland, 1999), pp. 42-53
- G. Sathe, S. Sarawagi, 2001. Intelligent Rollups in Multidimensional OLAP Data. VLDB (Roma, Italy 2001), pp.531-540

Text synthesis from query results

- A. Simitsis, G. Koutrika, Y. Alexandrakis, Y.E. Ioannidis, 2008. Synthesizing structured text from logical database subsets. EDBT (Nantes, France, 2008) pp. 428-439, DOI=<http://doi.acm.org/10.1145/1353343.1353396>

Formalities

OLAP Model

- ▶ We base our approach on an OLAP model that involves
 - Dimensions, defined as lattices of dimension levels
 - Ancestor functions, (in the form of $\text{anc}_{L_1}^{L_2}$) mapping values between related levels of a dimension
 - Detailed data sets, practically modeling fact tables at the lowest granule of information
 - Cubes, defined as aggregations over detailed data sets

What is Cube?

- ▶ A primary Cube C is described as

$$C = (DS^0, \Phi, [L_1, \dots, L_n, M_1, \dots, M_m], [agg_1(M_1^0), \dots, agg_1(M_m^0)])$$

- DS^0 is a detailed dataset over the schema
- Φ is a detailed selection condition
 - Φ analyzed as $\varphi_1 \wedge \dots \wedge \varphi_k$
 - φ_i is $D_i.L_j = value_i$
- L_1, \dots, L_n are levels such that $L_i < L_{i+1}$, $1 \leq i \leq n$.
- M_1, \dots, M_m are measures
- $agg_i \in \{max, min, sum, count, average\}$, $1 \leq i \leq m$

Cube Query

- ▶ A cube query Q can be considered as

$$Q = (DS^0, \Sigma, \Gamma, \gamma(M))$$

- ▶ where:

- Σ is a conjunction of dimensional restrictions of the form
- Γ is a set of grouper dimensional level
- $\gamma(M)$ is an aggregate function applied to the measure of the cube

Cube Query

- ▶ In our approach we assume that the user submit cube queries which denote as:
 - $q=(DS^0, \varphi_1 \wedge \cdots \wedge \varphi_k, [L_\alpha, L_\beta], \text{agg}(M))$
- ▶ Example:

$q=(A, W.L_2 = \text{'With-Pay'} \wedge E.L_3 = \text{'Post-Sec'}, [W.L_1, E.L_2], \text{avg}(\text{Hrs}))$

Cube Query to SQL Query

- ▶ In general case :

```
SELECT  $L_1, \dots, L_n, agg_1(M_1^0), \dots, agg_1(M_1^0)]$   
FROM  $DS^0$  INNER JOIN  $D_1, \dots$  INNER JOIN  $D_n$   
WHERE  $\phi$   
GROUP BY  $L_1, \dots, L_n$ 
```

- ▶ Example for our case:

```
SELECT W.L1, E.L2, AVG(Hrs)  
FROM A  
INNER JOIN W ON A.W=W.L0  
INNER JOIN E ON A.E=E.L0  
WHERE W.L2 = 'With-Pay' AND E.L3 = 'Post-Sec'  
GROUP BY W.L1, E.L2
```

Method Internals

Act I – Our Definition

- ▶ We introduce **two marginal sibling queries**, one for each aggregator.

- ▶ Formally, given an original query:

$$q = (DS^0, \varphi_1 \wedge \cdots \wedge \varphi_k, [L_\alpha, L_\beta], \text{agg}(M))$$

- ▶ Its two marginal sibling queries are:

1. $q^s = (DS^0, \varphi_1 \wedge \cdots \wedge \varphi_\chi^* \wedge \cdots \wedge \varphi_k, [L_\alpha, L_\chi], \text{agg}(M))$

2. $q^s = (DS^0, \varphi_1 \wedge \cdots \wedge \varphi_\chi^* \wedge \cdots \wedge \varphi_k, [L_\chi, L_\beta], \text{agg}(M))$

- $\varphi_\chi^*: L_{x+1} = \text{anc}_{L_x}^{L_{x+1}}(v)$

Act I – Query Example

▶ Original Query

- $q=(DS^0, W.L_2 = \text{'With-Pay'} \wedge E.L_3 = \text{'Post-Sec'}, [W.L_1, E.L_2], \text{avg(Hrs)})$

▶ Sibling Queries:

1. $q=(DS^0, W.L_2 = \text{'With-Pay'} \wedge E.L_4 = \text{'All'}, [W.L_1, E.L_3], \text{avg(Hrs)})$
2. $q=(DS^0, W.L_3 = \text{'All'} \wedge E.L_3 = \text{'Post-Sec'}, [W.L_2, E.L_2], \text{avg(Hrs)})$

Act I – How produce it?

- ▶ We define a sibling query as a query with a single difference to the original:
 - Instead of an atomic selection formula $L_i=v_i$, the sibling query contains a formula of the form $L_i \in \text{children}(\text{parent}(v_i))$.

- ▶ Formally, given an original query

$$q = (DS^0, \varphi_1 \wedge \cdots \wedge \varphi_k, [L_\alpha, L_\beta], \text{agg}(M))$$

- ▶ A new query q^s is a sibling query if is of the form

$$q^s = (DS^0, \varphi_1 \wedge \cdots \wedge \varphi_\chi^* \wedge \cdots \wedge \varphi_k, [L_\alpha, L_\beta], \text{agg}(M))$$

$$\bullet \varphi_\chi^*: L_{x+1} = \text{anc}_{L_x}^{L_{x+1}}(v)$$

Act II – Query Example

▶ Original Query

- $q=(DS^0, W.L_2 = 'With-Pay' \wedge E.L_3 = 'Post-Sec', [W.L_1, E.L_2], avg(Hrs))$

▶ Drill in Queries for work dimension:

1. $q=(DS^0, W.L_1 = 'Gov' \wedge E.L_3 = 'Post-Sec', [W.L_0, E.L_2], avg(Hrs))$

2. $q=(DS^0, W.L_1 = 'Private' \wedge E.L_3 = 'Post-Sec', [W.L_0, E.L_2], avg(Hrs))$

3. $q=(DS^0, W.L_1 = 'Self-emp' \wedge E.L_3 = 'Post-Sec', [W.L_0, E.L_2], avg(Hrs))$

For Education dimension: similarly

Act II- How produce it?

- ▶ Assume a cube query and its result, visualized as a 2D matrix.
- ▶ For each cell c of this result is characterized by the following cube query:
 - $q^c = (DS^0, \phi_1 \wedge \dots \wedge \phi_k \wedge \phi_c, [L_\alpha, L_\beta], \text{agg}(M))$
 - $\varphi_c : L_\alpha = v_a^c \wedge L_\beta = v_\beta^c$

Act II- How produce it?

- ▶ For each of the aggregator dimensions, we can generate a set of **explanatory drill in queries**, one per value in the original result:

1. $q_a^s = (DS^0, \phi_1 \wedge \dots \wedge \phi_k \wedge \phi, [L_{\alpha-1}, L_\beta], \text{agg}(M))$,

2. $q_\beta^s = (DS^0, \varphi_1 \wedge \dots \wedge \varphi_k \wedge \varphi_c, [L_\alpha, L_{\beta-1}], \text{agg}(M))$

• $\varphi_c : L_\alpha = v_a^c \wedge L_\beta = v_\beta^c$

Our Algorithm

Algorithm Construct Operational Act

Input: the original query over the appropriate database

Output: a set of an act's episodes fully computed

1. Create the necessary objects (act, episodes, tasks, subtasks) appropriately linked to each other
2. Construct the necessary queries for all the subtasks of the Act, execute them, and organize the result as a set of aggregated cells (each including its coordinates, its measure and the number of its generating detailed tuples)
3. For each episode
 - Calculate the visual presentation of cells
 - Calculate the cells' highlights
 - Produce the text based on the highlights
 - Produce the audio based on the text

Experiments and Results

Experimental setup

- Adult dataset referring to data from 1994 USA census
 - Has 7 dimension Age, Native Country, Education, Occupation, Marital status, Work class, and Race.
 - One Measure : work hours per week
- Machine Setup :
 - Running Windows 7
 - Intel Core Duo CPU at 2.50GHz,
 - 3GB main memory.

Experiments

Time breakdown(msec) per Act

