#### Schema Evolution and Gravitation to Rigidity: a tale of calmness in the lives of structured data

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http://www.cs.uoi.gr/~pvassil/projects/schemaBiographies/



"It is not the strongest of the species that survives, nor the most intelligent, but the one most responsive to *change*."

-Charles Darwin, 1809



Nicomachean Ethics, Book VII, Aristotle

		Correction	Enhancement
SWEBOK Maintenance	Proactive	Preventive	Perfective
	Reactive	Corrective	Adaptive

- Preventive maintenance: modification of a software product after delivery to detect and correct latent faults in the software product before they become operational faults.
- Corrective maintenance: reactive modification (or repairs) of a software product performed after delivery to correct discovered problems.
- Perfective maintenance: modification of a software product after delivery to provide enhancements for users, improvement of program documentation, and recoding to improve software performance, maintainability, or other software attributes.
- Adaptive maintenance: modification of a software product performed after delivery to keep a software product usable in a changed or changing environment.

## Database Evolution: why and what

- All software systems and, thus, both the databases themselves and applications built around databases are dynamic environments and can evolve due
  - Changes of requirements
  - Internal **restructuring** due to performance reasons
  - migration to / integration with another system
  - ...
- Database evolution further concerns
  - changes in the **operational environment** of the database
  - changes in the content (**data**) of the databases as time passes by
  - changes in the internal structure, or **schema**, of the database

# Why is (schema) evolution so important?

- Software and DB maintenance makes up for at least 50% of all resources spent in a project.
- Dependency magnets
  - Databases are rarely stand-alone: typically, an entire ecosystem of applications is structured around them =>
  - Typically, development waits till the "db backbone" is stable and applications are "safely" build on top of it, as...
  - ... changes in the schema can impact a large (typically, not traced) number of surrounding applications, without explicit identification of the impact & can cause several (parts of) different applications to crash, slow down, or miss data, causing the need for emergency repairing

### Evolving data-intensive ecosystem



### Evolving data-intensive ecosystem



The impact can be syntactical (causing crashes), semantic (causing info loss or inconsistencies) and related to the performance

## The impact of evolution

- Syntactic: scripts & reports simply crash
- **Semantic**: views and applications can become inconsistent or information losing
- Performance: can vary a lot



## We would really love to...

#### **Engineering goals**

- ... "design for evolution" and minimize the impact of evolution to the surrounding applications by introducing appropriate mechanisms in our DBMS's, applying design patterns & avoiding anti-patterns in both the db and the code in a way that insulates applications from unwanted schema change impacts
- ... plan in advance administration and perfective maintenance tasks and resources, instead of responding to emergencies

**Scientific goals** 

- ... (btw) detect & assess if there exist fundamental flaws in our Paradigms (like the relational model or the development of dataintensive applications)
- ... (with your permission) satisfy the scientific curiosity on gaining more knowledge on how things work
- ... but first, ...

... but, first, we must answer this:

WHAT ARE THE "LAWS" OF DATABASE SCHEMA EVOLUTION?



## Long term research goals

- Are there any "invariant properties" (e.g., patterns of repeating behavior) on the way database schemata change?
- Is there a theory / model to explain them?
- Can we exploit findings to engineer dataintensive ecosystems that withstand change gracefully?

http://www.cs.uoi.gr/~pvassil/projects/schemaBiographies/

# Do we know the mechanics of schema evolution?

- Historically, nobody from the research community had access
   + the right to publish to version histories of database
   schemata
- <u>Open source tools internally hosting databases have changed</u> this landscape, so...
- ... we are now presented with the opportunity to study the version histories of such "open source databases"



CLOSET

## Our take on the problem

- Collected version histories for the schemata of 8 open-source projects
  - CMS's: MediaWiki, TYPO3, Coppermine, phpBB, OpenCart
  - Physics: ATLAS Trigger --- Bio: Ensemble, BioSQL
- Preprocessed them to be parsable by our HECATE schema comparison tool and exported the transitions between each two subsequent versions and measures for them (size, growth, changes)
- Exploratory search where we statistically studied / mined these measures, to extract patterns & regularities for the lives of tables
- Web:

http://www.cs.uoi.gr/~pvassil/projects/schemaBiographies/

 Data and code available at: <u>https://github.com/DAINTINESS-Group</u>

## Scope of our studies

#### • Scope:

- databases being part of open-source software (and not proprietary ones)
- long history
- we work only with changes at the logical schema level (and ignore physical-level changes like index creation or change of storage engine)
- We encompass datasets with different domains ([A]: physics, [B]: biomedical, [C]: CMS's), amount of growth (shade: high, med, low) & schema size
- We should be very careful to not overgeneralize findings to proprietary databases or physical schemata!

FoSS Dataset	Versio ns	Lifetime	Tables @ Start	Tables @ End
ATLAS Trigger [A]	84	<mark>2</mark> Y, 7 M, 2 D	56	73
BioSQL [B]	46	<b>10 Y</b> , 6 M, 19 D	21	28
Coppermine [C]	117	<mark>8 Y</mark> , 6 M, 2 D	8	22
Ensembl [B]	528	<b>13 Y</b> , 3 M, 15 D	17	75
MediaWiki [C]	322	<mark>8 Y</mark> , 10 M, 6 D	17	50
OpenCart [C]	164	<b>4 Y</b> , 4 M, 3 D	46	114
phpBB [C]	133	<mark>6 Y</mark> , 7 M, 10 D	61	65
ТҮРОЗ [С]	97	<mark>8 Y</mark> , 11 M, 0 D	<b>10</b>	<b>23</b> 4

## How does the schema size evolve?

Normal

Explanat

mat

#### Outline

Paste

0 rando

5 trackbac validate

user restric

7 transcach

RR searchinde

9 objectcache

IF FO

Clipboa

AG53

- Schema size evolution
- Foreign Key Evolution
- Table Evolution
- Closing Remarks

#### Input: schema histories from github/sourceforge/...

**Output:** properties & patterns on the evolution of schema size (no. tables)

- Not covered here:
  - Growth patterns
  - Lehman laws & schema evolution



#### Schema Size (relations)



## Highlights of Schema Size Evolution



- Overall increase in size
- Periods of increase, esp. at beginning and after large drops
- **Drops**: sudden and steep (in short duration)
- Large periods of stability!
  - Unlike traditional S/W, db's are dependency magnets...



**Opencart: #Tables over time** Mwiki: #Tables over time Oct-09 May-10 Dec-10 Jul-11 Feb-12 Sep-12 Mar-09 Apr-13



Growth over time Calmness periods Ensembl: #Tables over time



Increase both slow (mostly) and abrupt Occasional abrupt drops (maintenance)



#### **Opencart: #Tables over time**



# Zipfian model in the distribution of growth frequencies





[With exceptions]

Density: <u>focused maintenance</u> effort <u>Progressive cooling</u>: early –maintenance density >> later stages Several spikes, many <u>zero-change periods/versions</u>

## How do foreign keys evolve?

Normal

Expla

mat

#### Outline

Paste

IF FO

Clipboa AG53

- Schema size evolution
- Foreign Key Evolution
- Table Evolution
- Closing Remarks

#### Input: schema histories from github/sourceforge/...

schema size @ end avg schema size 171 2.00 4.00 3.00 4.00 2.00 2.00 rande 5.00 2.00 1.6% 1.0 2.11 16.00 3.84 6.00 2.6% trackbad 6.37 0.05 4,0% 2.3 8.00 0.25 25.0% 12.00 22 246 2.00 5.6% 10.98 2.00 4.00 1.00 3.00

http://www.cs.uoi.gr/~pvassil/publications/2017\_ER/

**Output:** properties & patterns on the evolution of foreign keys

- Mainly patterns on:
  - Size

5200

5150

100

1 30

3 15

mwiki: updates / size

 When FK Births & Deaths

mwiki: duration / size

15

Quiet Dead

Active, Dea

Rigid

schema size@birth

... unexpected results...

200

5 150

100

50



III Active Su



## Evolution of Tables & FK's

- Tables grow in all cases
  (known from previous
  research) with periods
  of slow growth,
  calmness, spikes of
  extension, and
  occasional cleanups
- Foreign Keys are treated with different mentalities. 3 families:
  - Scientific
  - Comp. Toolkits
  - CMS's

## Evolution of Tables & FK's: Scientific projects



- Tables and FKS grow in synch, in both cases
- Growth comes with expansion periods, shrinkage actions, and periods of calmness in terms of both tables and foreign keys.

## Evolution of Tables & FK's: Computational Resource Toolkits



- Tables and FKS grow little and slowly; for Castor, not exactly in sync
- Castor: observe how scarce FK's are (too few tables come with FK's, see vertical axis)

## Evolution of Tables & FK's: Content Management Systems (CMS's)



- FK scarcity: really big at Slashcode, moderate at Zabbix
- Slashcode started <u>without</u> foreign keys at all; 1<sup>st</sup> set of FK's in v. 74.
   Zabbix seems to show a certain degree of syncronized growth
- Yet, ... both CMS's end up with no FK's!! -> see next

# What an unpleasant surprise: developers can resort in full removal of foreign keys!



- Slashcode: there is a clear phase of progressive removal
- Zabbix: abrupt removal of almost the entire set of foreign keys in a single transition. We have <u>no knowledge on why</u> this happened, & it is <u>unexpected</u> based on how FK's had been treated till then...

#### Slashcode: the disappearing FK's



"Commented-out foreign keys are ones which currently cannot be used because they refer to a primary key which is NOT NULL AUTO INCREMENT and the child's key either has a default value which would be invalid for an auto increment field, typically NOT NULL DEFAULT '0'.

Or, in some cases, the primary key is e.g. VARCHAR(20) NOT NULL and the child's key will be VARCHAR(20). The possibility of NULLs negates the ability to add a foreign key. <= That's my current theory, but it doesn't explain why discussions.topic SMALLINT UNSIGNED NOT NULL DEFAULT '0' is able to be foreign-keyed to topics.tid SMALLINT UNSIGNED NOT NULL AUTO INCREMENT"

"Stories is now InnoDB and these other tables are still MyISAM, so no foreign keys between them."

"This doesn't work, makes createStory die. These don't work, should check why..."

"This doesn't work, since in the install pollquestions is populated before users, alphabetically"

"This doesn't work, since discussion may be 0."

1<sup>st</sup> massive foreign key removal (rev 1.120), 22 FK's deleted.

2<sup>nd</sup> massive deletion (rev 1.151), 10 FK's deleted

3<sup>rd</sup> deletion (rev 1.174), 3 FK's deleted

4<sup>th</sup> deletion (rev 1.189) 1 FK deleted

5<sup>th</sup> deletion (rev 1.201) 1 FK<sup>2</sup>deleted

#### Slashcode: what did the comments say?

- The main problem seems to be the difficulty of developers with the tuning and handling of both foreign and primary keys.
- Sometimes <u>difficulties are hard</u> -- e.g., different storage engines, typically due to performance reasons
- Some difficulties are complicated <u>due to technicalities</u> like autonumbering
- Sometimes <u>fixes could be found with some effort (e.g.,</u> changing the order of table population, or using numeric data types for primary keys, or inserting some "goalkeeper" values at FK target table)

### Scarcity of Foreign keys

A 2013 collection of schema histories, lists 21 data sets,
 -- some have more than one target DBMS variants.

<pre>\$ cd RESEARCH/Github/EvolutionDatasets</pre>										
\$ ls -d * */*										
CERN	CMS's/Coppermine	CMS's/XOOPS	Med							
CERN/Atlas	CMS's/DekiWiki	CMS's/Zabbix	Med/Ensembl							
CERN/CASTOR	CMS's/Joomla 1.5	CMS's/e107	Med/biosql							
CERN/DQ2	CMS's/NucleusCMS	CMS's/opencart	README.md							
CERN/DRAC	CMS's/SlashCode	CMS's/phpBB								
CERN/EGEE	CMS's/TikiWiki	CMS's/phpwiki								
CMS's	CMS's/Typo3	CMS's/wikimedia								

How many data sets contain foreign keys?

Try this (also backed by manual sampling):

grep -rl "FOREIGN" . >> ALL-FKs-by-grep.ascii
awk '{split(\$0,a,"/"); print a[2],a[3]}' ALL-FKs-by-grep.ascii
uniq

#### Scarcity of Foreign keys

#### - How many data sets, out of the 21, contain foreign keys?

CERN Atlas CERN CASTOR CERN EGEE CMS's SlashC CMS's Zabbix Med biosql

CERN DQ2 CERN DIRAC Med Ensembl

#### The 6 data sets reported here

**DQ2** (only in the mySQL, not in the Oracle version): FK's in 19 versions out of the 55. Starts with 2 FK's and ends with 1.

**DIRAC** (not in the production folder, only at python+mysql). 9 tables at first version, 15 tables at last version Starts with 10 FK's, ends with 8

**Ensembl**: not able to link FK DDL files to table evolution, yet

 - 9 out of the 21 data sets do (including 3 that are really small for harnessing valuable results, spec., Egee, DQ2, DIRAC) you're not welcome here...

#sorrynotsorry

# Foreign Key Evolution comes with different treatments:

- Sometimes, **FK's are treated as an integral part of the system**, and they are born and evicted along with table birth and eviction.
- Other times, **FK's are treated as a disposable add-on**: only a small subset of the tables involved in FK's; birth and eviction of FK's rarely performed in synch with their tables. If technical difficulties arise, it is possible to witness the **complete removal of FK's** from the schema.
- Another sign of concern is that in all the CMS' we collected, FK's are too scarce

To appear

in ER 2017

• More results in the paper: **stats**, **threats to validity**, and, the treatment of the **evolving schema as an evolving graph** 

http://www.cs.uoi.gr/~pvassil/publications/2017\_ER/

## How do individual tables evolve?

Normal

Check Cel

Dead: 10 Survi: 20 Bad

10

21

#### <u>Outline</u>

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AG53

Paste

- Schema size evolution
- Foreign Key Evolution
- Table Evolution
- Closing Remarks

# **Input:** schema histories from github/sourceforge/...

Output: properties & patterns on table properties (birth, duration, amt of change, ...)

#### **Highlights**

<u>4 patterns of evolution,</u> here we focus on two of them



9															
10	tableName	duration	birth	death	schema size@birth	schema size @ end	avg schema size	sum(updates)	count(updates)	ATU	UpdateRate	AvgUpdVolume	SizeScaleUp i	aad/Surviv	Activity di
11	/*\$wgDBPrefix*/protected_titles	1	171	171	7	7	7.00	0	0	0.00	0.0%		1.00	10	0
12	blobs	35	20	54	2	2	2.00	0	0	0.00	0.0%		1.00	10	0
13	brokenlinks	62	O	61	2	2	2.00	0	0	0.00	0.0%		1.00	10	0
14	concurrencycheck	1	317	317	4	4	4.00	0	0	0.00	0.0%		1.00	10	0
15	globalinterwiki	5	294	300	2	2	2.00	0	0	0.00	0.0%		1.00	10	0
16	globalnamespaces	3	294	300	5	3	3.00	0	0	0.00	0.0%		1.00	10	0
17	globaltemplatelinks	3	294	300	B	8	8.00	0	0	0.00	0.0%		1.00	10	0
18	groups	6	59	64	4	4	4.00	0	0	0.00	0.0%		1.00	10	0
19	imageredirects	1	175	175	2	2	2.00	0	0	0.00	0.0%		1.00	10	0
20	random	2	0	1	2	2	2.00	0	0	0.00	0.0%		1.00	10	0
21	math	282	0	281	5	5	5.00	3	1	0.01	0.496	3.0	1.00	10	1
22	links	62	0	61	2	2	2.00	1	1	0.02	1.6%	1.0	1.00	10	1
23	linksec	57	6	62	3	2	2.11	1	1	0.02	1.8%	1.0	0.67	10	1
24	cur	41	0	40	15	16	16.00	1	1	0.02	2.4%	1.0	1.00	10	1
25	group	25	34	58	3	4	3.84	1	1	0.04	4.0%	1.0	1.53	10	1
26	trackbacks	235	74	308	5	6	6.00	10	6	0.04	2.6%	1.7	1.20	10	1
27	validate	75	23	97	5	7	6.37	7	3	0.09	4.0%	2.3	1.40	10	1
28	recentlinkchanges	4	197	200	8	8	8.00	1	1	0.25	25.0%	1.0	1.00	10	1
29	user_restrictions	3	210	214	12	12	12.00	3	1	1.00	33.3%	3.0	1.00	10	1
30	user rights	36	29	64	2	2	2.00	6	2	0.17	5.6%	3.0	1.00	10	2
31	old	41	o	40	11	11	10.95	7	3	0.17	7.3%	2.3	1.00	10	z
32	config	39	284 -		2	2	2.00	0	0	0.00	0.0%		1.00	20	0
33	tag_summary	100	223 -		4	4	4.00	0	0	0.00	0.0%		1.00	20	0
34	hitcounter	315	7 -		1	1	1.00	1	1	0.00	0.3%	1.0	1.00	20	1
35	externallinks	236	87 -		3	3	3.00	1	1	0.00	0.4%	1.0	1.00	20	1
36	text	282	41 -		3	3	3.00	2	2	0.01	0.7%	1.0	1.00	20	1
37	transcache	235	88 -		3	3	3.00	2	2	0.01	0,9%	1.0	1.00	20	1
38	searchindex	323	0 -		3	3	3.00	3	3	0.01	0.9%	1.0	1.00	20	1
39	objectcache	307	16 -		3	3	3.00	3	з	0.01	1.0%	1.0	1.00	20	1
40	user_properties	89	234 -		5	3	3.00	1	1	0.01	1.196	1.0	1.00	20	1
41	pagelinks	262	61 -		3	3	3.00	3	3	0.01	1.196	1.0	1.00	20	1

#### **Regularities on table change do exist**!





#### If you're wide, you survive





Top-changers typically live long, are early born, survive ...

... and they are not necessarily the widest ones in terms of schema size



For details: - ER 2015 - Inf. Sys. 2017 Progressive cooling: most change activity lies at the beginning of the db history
Void triangle: The few dead tables are typically quiet, early born, short lived, and quite often all three of them



# Longevity and update activity correlate !!

The few top-changers (in terms of avg trans. update – ATU)

- are long lived,
- typically come from the early versions of the database
- due to the combination of high ATU and duration => they have high total amount of updates, and,
- frequently survive!


#### mwiki: duration / birth



### Die young and suddenly

- There is a very large concentration of the deleted tables in a small range of newly born, quickly removed, with few or no updates...
  - .... resulting in very
    low numbers of
    removed tables with
    medium or long
    durations (empty
    triangle).





- CAiSE 2017

# ELECTROLYSIS PATTERN FOR TABLE ACTIVITIES For details:

http://www.cs.uoi.gr/~pvassil/publications/2017\_CAiSE\_Electrolysis

# The electrolysis pattern



- Dead tables demonstrate much shorter lifetimes than survivor ones,
- can be located at short or medium durations, and practically never at high durations.
- With few exceptions, the less active dead tables are, the higher the chance to reach shorter durations.
- Survivors expose the inverse behavior, i.e., mostly located at medium or high durations.
- The more active survivors are, the stronger they are attracted towards high durations, with a significant such inclination for the few active ones that cluster in very high durations.



0%

Year Range

0 4

ActiveSurv

QuietSurv

Rieidsury

Active Dead

Quiet Dead

100%

50%

0%

100%

50%

0%

Year Range

0 1 2

3

4

ActiveSur

Quiet Surv

ANG ASUN

Quiet

Rigid Dead

0

Year Range

1

2

5

6

pct for class

pct for class

# Electrolysis as a heatmap showing the extreme bias between dead and survivor tables



- For each *LifeAndDeath* value, and for each duration range of 5% of the database lifetime, we computed the percentage of tables (over the total of the data set) whose duration falls within this range.
- We removed cells that corresponded to only one data set

The resulting heatmap shows the polarization in colors: brighter color signifies higher percentage of the population

# Gravitation to Rigidity



- Although the majority of survivor tables are in the quiet class, we can quite emphatically say that it is the absence of evolution that dominates!
  - Survivors vastly outnumber removed tables.
  - Similarly, rigid tables outnumber the active ones, both in the survival and, in particular, in the dead class.
  - Schema size is rarely resized, and only in survivors (not in the paper).
  - Active tables are few and do not seem to be born in other but early phases of the database lifetime.
- Evidently, not only survival is also stronger than removal, but **rigidity is also stronger a force than variability** and the combination of the two forces further lowers the amount of change in the life of a database schema.

# Summarizing...



- Yes, we can indeed find patterns in the lives of tables, during schema evolution!
- <u>Survivors</u>, mostly <u>long-lived</u> (esp. active ones) and <u>quietly active</u> are <u>radically different</u> than <u>dead</u> tables, being mostly <u>short-lived</u> and <u>rigid</u>!
- Gravitation to rigidity rules: we see more absence than presence of schema evolution!



Also studied [not part of the paper]: year of birth, schema size, schema resizing

http://www.cs.uoi.gr/~pvassil/publications/2017\_CAiSE\_Electrolysis

#### <u>Outline</u>

- Schema size evolution
- Foreign Key Evolution
- Table Evolution
- Closing Remarks

Where we stand

Open issues

... and discussions ...

## **CLOSING REMARKS**



### Where we stand

We have a first understanding of ...

- gravitation to rigidity, i.e., the mechanics of schema non-evolution for FoSS ecosystems
- schemata growing, changed in focused periods of maintenance and progressively "cooling" down
- patterns relating to how tables change, given their size, update behavior, time of birth, ...
- foreign key families of treatment, absence & removals

To probe further (code, data, details, presentations, ...) <a href="http://www.cs.uoi.gr/~pvassil/projects/schemaBiographies/">http://www.cs.uoi.gr/~pvassil/projects/schemaBiographies/</a>

#### **Gravitation to rigidity:**

- Long calmness, low+focused growth
- Empty triangle, inverse Gamma, electrolysis

More absence than presence of evo!



#### Individual Tables











#### **Foreign Keys**



# Where to go from here...

- More studies, by more groups, on more data, to verify / disprove patterns & find new ones
- More tools and techniques to fully automate processing
- Weather Forecast: given the history and the state of a database, predict subsequent events
- Engineer for evolution: To absorb change gracefully we can try to (i) alter db design and DDL; (ii) encapsulate the database via a "stable" API; ...

# How does schema evolution relate to the surrounding software?

- Which parts of the surrounding data-intensive software app's are most sensitive to evolution?
  - Metrics for sensitivity to evolution?
  - Visualization of the architecture & evolution impact

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- Automation of the reaction to changes
  - self-monitoring
  - impact prediction
  - auto-regulation (policy determination)
  - self-repairing

http://www.cs.uoi.gr/~pvassil/projects/hecataeus/

#### daintiness DAta INTensive Information EcoSystemS Group https://github.com/DAINTINESS-Group

Repositories	🕀 People 🔫	🖞 Teams 4	🔅 Settings
Filters - Q Find	d a repository		
EvolutionDat	asets		
¥ forked from giskou/Ev	olutionDatasets		
Updated on 31 Jul			
Hecate			

¥ forked from giskou/Hecate

Diff visualization between 2 SQL schemas

Updated on 2 Apr

#### **Everything HAS TO BE online!**

CO

C

We are happy to invite you to reuse / test / disprove /... all our code, data and results!

#### To probe further (code, data, details, presentations, ...) <a href="http://www.cs.uoi.gr/~pvassil/projects/schemaBiographies/">http://www.cs.uoi.gr/~pvassil/projects/schemaBiographies/</a>

# Thank you!

- Yes, we have the data and the tools to find patterns of schema evolution both for the entire schema and for individual parts of it!
- Gravitation to rigidity rules: we see more absence than presence of schema evolution!
- Many opportunities to exploit data, code and results for research on more studies, design and visualization of systems

To probe further (code, data, details, presentations, ...) <a href="http://www.cs.uoi.gr/~pvassil/projects/schemaBiographies/">http://www.cs.uoi.gr/~pvassil/projects/schemaBiographies/</a>

### **AUXILIARY SLIDES**

## Embedded queries in the past [Maule+08] ...

```
10
    public static IEnumerable<Experiment> Q1(DateTime d) {
11
     DBParams dbParams = new DBParams();
12
    DBRecordSet queryResult;
13
     List<Experiment> exps = new List<Experiment>();
14
15
     dbParams.Add("@ExpDate", d);
16
17
     queryResult = QueryRunner.Run(
18
      "SELECT Experiments.Name, Experiments.ExperimentId"+
      " FROM Experiments"+
19
20
      " WHERE Experiments.Date={@ExpDate}",
      dbParams);
21
22
     while (queryResult.MoveNext()) {
23
      exps.Add(new Experiment(queryResult.Record));
24
25
     }
26
27
     return exps;
28
    ł
```

# ... nowadays, to be complemented with API-based db access (Drupal)

```
function _profile_get_fields($category, $register = FALSE) {
 $query = db_select('profile_field');
 if ($register) {
  $query->condition('register', 1);
 else {
  $query->condition('category', db_like($category), 'LIKE');
 7
 if (!user_access('administer users')) {
  $query->condition('visibility', PROFILE_HIDDEN, '<>');
 7
 return $query
   ->fields('profile_field')
   ->orderBy('category', 'ASC')
   ->orderBy('weight', 'ASC')
   ->execute();
ł
```

# Abstract coupling example from my SW Dev course

Interface as a contract



# Put it all online!!

#### https://github.com/DAINTINESS-Group/

My web page

http://www.cs.uoi.gr/~pvassil/

has links to ... DB Schema Evolution Papers, **Data sets, Code, Results** projects/schemaBiographies/

... and to ...

#### Tools for handling Evolution (Hecataeus)

projects/hecataeus/

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Updated 15 days	ago			
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Diff visualization	between 2 SQL schemas			
Updated on 2 Apr				
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	olution of the tables of a database schema as parallel lives			

Visualizes the evolution of the tables of a database schema as parallel li Updated on 2 Apr

#### Hecataeus

Y forked from pmanousis/Hecataeus Database evolution what-if analysis tool Updated on 24 Oct 2014 Java 🖈 0 🖗 2

## SCOPE OF THE STUDY && VALIDITY CONSIDERATIONS

### Datasets

https://github.com/DAINTINESS-Group/EvolutionDatasets

- Content management Systems
  - MediaWiki, TYPO3, Coppermine, phpBB, OpenCart
- Medical Databases
  - Ensemble, BioSQL
- Scientific
  - ATLAS Trigger

### Data sets

Dataset	Versi ons	Lifetime	Table s Start	Table s End	Attribut es Start	Attribut es End	Commit s per Day	% commits with change	Repository URL
ATLAS Trigger	84	2 Y, 7 M, 2 D	56	73	709	858	0,089	82%	http://atdaq-sw.cern.ch/cgi-bin/viewcvs- atlas.cgi/offline/Trigger/TrigConfiguration/TrigDb/share/sql/com bined_schema.sql
BioSQL	46	10 Y, 6 M, 19 D	21	28	74	129	0,012	63%	https://github.com/biosql/biosql/blob/master/sql/biosqldb- mysql.sql
Coppermine	117	8 Y, 6 M, 2 D	8	22	87	169	0,038	50%	http://sourceforge.net/p/coppermine/code/8581/tree/trunk/cpg 1.5.x/sql/schema.sql
Ensembl	528	13 Y, 3 M, 15 D	17	75	75	486	0,109	60%	http://cvs.sanger.ac.uk/cgi- bin/viewvc.cgi/ensembl/sql/table.sql?root=ensembl&view=log
MediaWiki	322	8 Y, 10 M, 6 D	17	50	100	318	0,100	59%	https://svn.wikimedia.org/viewvc/mediawiki/trunk/phase3/main tenance/tables.sql?view=log
OpenCart	164	4 Y, 4 M, 3 D	46	114	292	731	0,104	47%	https://github.com/opencart/opencart/blob/master/upload/inst all/opencart.sql
phpBB	133	6 Y, 7 M, 10 D	61	65	611	565	0,055	82%	https://github.com/phpbb/phpbb3/blob/develop/phpBB/install/ schemas/mysql_41_schema.sql
ТҮРОЗ	97	8 Y, 11 M, 0 D	10	23	122	414	0,030	76%	https://git.typo3.org/Packages/TYPO3.CMS.git/history/TYPO3_6- 0:/t3lib/stddb/tables.sql

# Scope of the study

#### • Scope:

- databases being part of open-source software (and not proprietary ones)
- long history
- we work only with changes at the logical schema level (and ignore physical-level changes like index creation or change of storage engine)
- We encompass datasets with different domains ([A]: physics, [B]: biomedical, [C]: CMS's), amount of growth (shade: high, med, low) & schema size
- We should be very careful to not overgeneralize findings to proprietary databases or physical schemata!

FoSS Dataset	Versio ns	Lifetime	Tables @ Start	Tables @ End
ATLAS Trigger [A]	84	<mark>2 Y</mark> , 7 M, 2 D	56	73
BioSQL [B]	46	<b>10 Y</b> , 6 M, 19 D	21	28
Coppermine [C]	117	<mark>8 Y</mark> , 6 M, 2 D	8	22
Ensembl [B]	528	<b>13 Y</b> , 3 M, 15 D	17	75
MediaWiki [C]	322	<mark>8 Y</mark> , 10 M, 6 D	17	50
OpenCart [C]	164	<mark>4 Y</mark> , 4 M, 3 D	46	114
phpBB [C]	133	<mark>6 Y</mark> , 7 M, 10 D	61	65
ТҮРОЗ [С]	97	<mark>8 Y</mark> , 11 M, 0 D	<b>10</b> 5	<b>23</b> 9

## Hecate: SQL schema diff extractor

- Parses DDL files
- Creates a model for the parsed SQL elements
- Compares two versions of the same schema
- Reports on the diff performed with a variety of metrics
- Exports the transitions that occurred in XML format

https://github.com/DAINTINESS-Group/Hecate

## Hecate: SQL schema diff extractor

Name	Туре	📉 Name	Туре
/ 🐉 rev_001284.sql		🔺 🔻 🜮 rev_113110.sql	·
🔻 🛄 archive		🔹 🛄 archive	
📒 ar_comment	TINYBLOB	ar_comment	TINYBLOB
📒 ar_flags	TINYBLOB	📃 ar_deleted	TINYINTUNSIGNED
🔲 ar_minor_edit	TINYINT(1)	🔲 ar_flags	TINYBLOB
ar_namespace	TINYINT(2)	🔲 ar_len	INTUNSIGNED
🔲 ar_text	MEDIUMTEXT	ar_minor_edit	TINYINT
🔲 ar_timestamp	CHAR(14)	🔲 ar_namespace	INT
📒 ar_title	VARCHAR(255)	📃 ar_page_id	INTUNSIGNED
🔲 ar_user	INT(5)	📃 ar_parent_id	INTUNSIGNED
📒 ar_user_text	VARCHAR(255)	📃 ar_rev_id	INTUNSIGNED
🔹 🔜 brokenlinks		🔲 ar_shal	VARBINARY(32)
🔲 bl_from	INT(8)	🔲 ar_text	MEDIUMBLOB
📃 bl_to	VARCHAR(255)	🔲 ar_text_id	INTUNSIGNED
🕨 🔜 cur		🔲 ar_timestamp	BINARY(14)
🕨 🛄 image		🔲 ar_title	VARCHAR(255)
🕨 🛄 imagelinks		🔲 ar_user	INTUNSIGNED
ipblocks		🔲 ar_user_text	VARCHAR(255)
🕨 📃 links		category	
🕨 🔜 math		🕨 📄 categorylinks	
🕨 📃 old		change_tag	
oldimage		🕨 🛄 config	
🕨 🔜 random		🕨 🔛 external_user	
recentchanges		🕨 🛄 externallinks	
searchindex		🕨 🛄 filearchive	
site_stats		🕨 🛄 hitcounter	
user		🔎 🕨 🛄 image	
user_newtalk		🚽 🕨 🛄 imagelinks	

#### https://github.com/DAINTINESS-Group/Hecate

Can we generalize out findings broadly?

# **External validity**

- We perform an **exploratory study to observe frequently occurring phenomena** within the scope of the aforementioned population
- Are our data sets representative enough? Is it possible that the observed behaviors are caused by sui-generis characteristics of the studied data sets?
  - Yes: we believe we have a good population definition & we abide by it
  - Yes: we believe we have a large number of databases, from a variety of domains with different profiles, that seem to give fairly consistent answers to our research questions (behavior deviations are mostly related to the maturity of the database and not to its application area).
  - Yes: we believe we have a good data extraction and measurement process without interference / selection / ... of the input from our part
  - Maybe: unclear when the number of studied databases is large enough to declare the general application of a pattern as "universal".



# **External validity**

Can we generalize out findings broadly?

- Understanding the represented population
  - Precision: all our data sets belong to the specified population
  - Definition Completeness: no missing property that we knowledgably omit to report
  - FoSS has an inherent way of maintenance and evolution
- Representativeness of selected datasets
  - Data sets come from 3 categories of FoSS (CMS / Biomedical / Physics)
  - They have different size and growth volumes
  - Results are fairly consistent both in our ER'15 and our CAiSE'14 papers
- Treatment of data
  - We have tested our "Delta Extractor", Hecate, to parse the input correctly & adapted it during its development; the parser is not a full-blown SQL parser, but robust to ignore parts unknown to it
  - A handful of cases where adapted in the Coppermine to avoid overcomplicating the parser; not a serious threat to validity; other than that we have not interfered with the input
  - Fully automated counting for the measures via Hecate

# Internal validity

- Can we confirm statements A=>B? No!
- Are there any spurious relationships? Maybe!
- Internal validity concerns the accuracy of causeeffect statements: "change in A => change in B"
- We are very careful to avoid making strong causation statements!
  - In some places, we just <u>hint</u> that we <u>suspect</u> the causes for a particular phenomenon, in some places in the text, but <u>we have no data, yet, to verify our gut-</u> <u>feeling</u>.
  - And yes, it is quite possible that our correlations hide cofounding variables.

# Is there a theory?

- Our study should be regarded as a pattern observer, rather than as a collection of laws, coming with their internal mechanics and architecture.
- It will take too many studies (to enlarge the representativeness even more) and more controlled experiments (in-depth excavation of cause-effect relationships) to produce a solid theory.
- It would be highly desirable if a clear set of requirements on the population definition, the breadth of study and the experimental protocol could be solidified by the scientific community (like e.g., the TREC benchmarks)
- ... and of course, there might be other suggestions on how to proceed...

### **RELATED WORK**



**Sjoberg @ IST 93**: 18 months study of a health system. 139% increase of #tables ; 274% increase of the #attributes

Changes in the code (on avg):

relation addition: 19 changes ; attribute additions: 2 changes
relation deletion : 59.5 changes; attribute deletions: 3.25 changes

An **inflating period** during construction where almost all changes were additions, and a **subsequent period** where additions and deletions where balanced.



Curino+ @ ICEIS08: Mediawiki for 4.5 years 100% increase in the number of tables 142% in the number of attributes.

45% of changes do not affect the information capacity of the schema (but are rather index adjustments, documentation, etc)



IWPSE09: Mozilla and Monotone (a version control system) Many ways to be out of synch between code and evolving db schema

ICDEW11: Firefox, Monotone, Biblioteq (catalogue man.), Vienna (RSS) Similar pct of changes with previous work Frequency and timing analysis: **db schemata tend to stabilize over time**, as there is more change at the beginning of their history, but seem to converge to a relatively fixed structure later



Qiu,Li,Su@ FSE 2013: 10 (!) database schemata studied. Change is focused both (a) with respect to time and (b) with respect to the tables who change.

**Timing**: 7 out of 10 databases reached 60% of their schema size within 20% of their early lifetime.

Change is frequent in the early stages of the databases, with inflationary characteristics; then, the schema evolution process calms down.

**Tables that change**: 40% of tables do not undergo any change at all, and 60%-90% of changes pertain to 20% of the tables (in other words, 80% of the tables live quiet lives). The most frequently modified tables attract 80% of the changes.

Sjoberg IST 93	Curino+ ICEIS08	Univ. Ri <sup>v</sup> IWPSE09,		Qiu,Li,Su FSE'13		Univ. Ioannina CAiSE14, ER15	
1993	2008	2009	2011	2013	2014	2015	

#### Qiu,Li,Su@ FSE 2013: Code and db co-evolution, not always in synch.

- Code and db changed in the same revision: 50.67% occasions
- Code change was in a previous/subsequent version than the one where the database schema change: 16.22% of occasions
- database changes not followed by code adaptation: 21.62% of occasions
- 11.49% of code changes were unrelated to the database evolution.

Each atomic change at the schema level is estimated to result in 10 -- 100 lines of application code been updated;

A valid db revision results in 100 -- 1000 lines of application code being updated


## **Timeline of empirical studies**

CAiSE14: DB level ER'15: Table level



.. What do we see if we observe the evolution of the entire schema?

http://www.cs.uoi.gr/~pvassil/publications/2014\_CAiSE/

- Skoulis, Vassiliadis, Zarras. Open-Source Databases: Within, Outside, or Beyond Lehman's Laws of Software Evolution? CAISE 2014
- Growing up with stability: How open-source relational databases evolve. Information Systems, Volume 53, October–November 2015

### SCHEMA EVOLUTION AND LEHMAN LAWS

## Exploratory search of the schema histories for patterns

Normal

Dead: 10 Survi: 20

Input: schema histories from github/sourceforge/... Raw material: details and stats on each table's life, as produced by our diff extractor, for all the 8 datasets

Paste

Output: properties &
patterns on schema
properties (size, growth,
<u>changes,)</u> that occur
frequently in our data sets
<u>Highlights</u>



 <u>Compliance to Lehman's</u> <u>laws</u>



9															
10		duration	birth	death	schema size@birth	schema size @ end	avg schema size	sum(updates)	count(updates)	ATU	UpdateRate	AvgUpdVolume	SizeScaleUp	aad/Surviv	v Activity d
11	/*SwgDBPrefix*/protected_titles	1	171	171	7	7	7.00	0	0	0.00	0.0%		1.00	10	0
12	blobs	35	20	54	2	2	2.00	0	0	0.00	0.0%		1.00	10	0
13	brokenlinks	62	0	61	2	2	2.00	0	0	0.00	0.0%		1.00	10	0
	concurrencycheck	1	317	317	4	4	4.00	0	0	0.00	0.0%		1.00	10	0
15	globalinterwiki	5	294	300	2	2	2.00	0	0	0.00	0.0%		1.00	10	0
16		3	294	300	3	3	3.00	0	0	0.00	0.0%		1.00	10	0
17	globaltemplatelinks	3	294	300	B	8	8.00	0	0	0.00	0.0%		1.00	10	0
18	groups	6	59	64	4	4	4.00	0	0	0.00	0.0%		1.00	10	0
19	imageredirects	1	175	175	2	2	2.00	0	0	0.00	0.0%		1.00	10	0
20	random	2	0	1	2	2	2.00	0	0	0.00	0.0%		1.00	10	0
21	math	282	0	281	5	5	5.00	3	1	0.01	0.496	3.0	1.00	10	1
22	links	62	0	61	2	2	2.00	1	1	0.02	1.6%	1.0	1.00	10	1
23	linksoc	57	6	62	3	2	2.11	1	1	0.02	1.8%	1.0	0.67	10	1
24	cur	41	0	40	16	16	16.00	1	1	0.02	2.4%	1.0	1.00	10	1
25	group	25	34	58	5	4	3.84	1	1	0.04	4.0%	1.0	1.53	10	1
26	trackbacks	235	74	308	5	6	6.00	10	6	0.04	2.6%	1.7	1.20	10	1
27	validate	75	23	97	5	7	6.37	7	3	0.09	4.0%	2.3	1.40	10	1
28	recentlinkchanges	4	197	200	8	8	8.00	1	1	0.25	25.0%	1.0	1.00	10	1
29	user_restrictions	3	210	214	12	12	12.00	3	1	1.00	33.3%	3.0	1.00	10	1
30	user_rights	36	29	64	2	2	2.00	6	2	0.17	5.6%	3.0	1.00	10	2
31	old	41	o	40	11	11	10.98	7	3	0.17	7.3%	2.3	1.00	10	2
32	config	39	284 -		2	2	2.00	0	0	0.00	0.0%		1.00	20	0
33	tag_summary	100	223 -		4	4	4.00	0	0	0.00	0.0%		1.00	20	0
34	hitcounter	315	7 -		1	1	1.00	1	1	0.00	0.3%	1.0	1.00	20	1
35	externallinks	235	87 -		3	3	3.00	1	1	0.00	0.4%	1.0	1.00	20	1
36	text	282	41 -		3	3	3.00	2	2	0.01	0.7%	1.0	1.00	20	1
37	transcache	235	88 -		3	3	3.00	2	2	0.01	0,9%	1.0	1.00	20	1
38	searchindex	323	0 -		3	3	3.00	3	3	0.01	0.9%	1.0	1.00	20	1
39	objectcache	307	16 -		3	3	3.00	3	з	0.01	1.0%	1.0	1.00	20	1
40	user_properties	89	234 -		3	3	3.00	1	1	0.01	1.1%	1.0	1.00	20	1
41	pagelinks	262	61 -		3	3	3.00	3	3	0.01	1.1%	1.0	1.00	20	1

### Schema Size (relations)



http://www.cs.uoi.gr/~pvassil/publications/2014\_CAiSE/

## Schema Size

- Overall increase in size
- Periods of increase, esp. at beginning and after large drops -> positive feedback
- Drops: sudden and steep (in short duration) -> negative feedback
- Large periods of stability!
  - Unlike traditional S/W, db's are dependency magnets...



**Opencart: #Tables over time** Mwiki: #Tables over time Oct-09 May-10 Dec-10 Jul-11 Feb-12 Sep-12 Mar-09 Apr-13



Growth over time Calmness periods Ensembl: #Tables over time



Increase both slow (mostly) and abrupt Occasional abrupt drops (maintenance)



#### **Opencart: #Tables over time**



#### Coppermine: #Tables over time







## Schema Growth (diff in #tables)



2014 CAiSE/

## Schema growth is small!

- Growth is bounded in small values!
- Zipfian distribution of growth values around 0
  - Predominantly: occurrences of zero growth; almost all deltas are bounded between [-2..2] tables
  - [0..2] tables slightly more popular => average value of growth slightly higher than 0
- No periods of continuous change; small spikes instead
- Due to perfective maintenance, we also have negative values of growth (less than the positive ones).
- Oscillations exist too: positive growth is followed with immediate negative growth or stability

# Zipfian model in the distribution of growth frequencies



## What happens after large changes?







Density: focused maintenance effort

**Progressive cooling**: early –maintenance density >> later stages Several spikes, many zero-change periods/versions

[With exceptions]

#### **#tables & heartbeat of changes over time**



## How do schemata evolve?

Schema size (#tables – also: #attributes) supports the assumption of a feedback mechanism

- Schema size grows over time; not continuously, but with bursts of concentrated effort
- Drops in schema size signify the existence of perfective maintenance
- Large periods of stability

#### Schema Growth (diff in size between subsequent versions) is small!!

- Growth is small, smaller than in typical software
- Average growth is close (slightly higher) to zero

#### **Gravitation to rigidity:**

- Large periods of stability
- Change frequency drops with time

For details: - CAiSE 2014 - Inf. Systems 2015

http://www.cs.uoi.gr/~pvassil/publications/2014\_CAiSE/



### THE FOUR PATTERNS



To probe further (code, data, details, presentations, ...) http://www.cs.uoi.gr/~pvassil/publications/2015 ER/

Activity	Top-changers (high ATU) are born early, live long, have large amt of update	
	<ul> <li>Inverse Γ :</li> <li>Top-changers: mostly at long durations</li> <li>Long duration: all kinds of change</li> </ul>	<ul> <li>Comet:</li> <li>Many updates: typically at medium schema size @ birth</li> <li>Large schema at birth: medium amount of updates</li> </ul>
Rigidity	<ul> <li>Inverse Γ :</li> <li>small duration → small change</li> <li>medium duration → small or medium change</li> </ul>	Comet: ~70% of tables $\in$ 10x10 narrow & quiet box
Survival	Γ : the majority of wide tables are created early on and survive	Γ : if you 're wide, you survive
	Heaven can wait for old-timers	
Death	Dead tables: quiet, early born, short-	
	lived, and quite often all three of them	

Duration & Birth

Schema size

## Statistical study of durations

#### Normalized Durations and their pct over #tables

- Short and long lived tables are practically equally proportioned
- Medium size durations are fewer than the rest!
- Long lived tables are surprisingly too many
  - in half the data sets they are the most populated group
  - in all but one data set they exceed 30%

		<u>Short</u>	<u>Medium</u>	Long	Long,	Max
	<u># tables</u>	Lived	Lived	Lived	<u>not max</u>	<b>Duration</b>
atlas	88	32%	14%	55%	5%	50%
biosql	45	31%	38%	31%	11%	20%
coppermine	23	0%	22%	78%	43%	35%
ensembl	155	55%	37%	8%	3%	5%
mwiki	71	46%	21%	32%	18%	14%
opencart	236	54%	9%	36%	36%	0%
phpBB	70	9%	10%	81%	0%	81%
typo3	32	34%	28%	38%	9%	28%
Overall	720	42%	20%	38%	18%	20%

Way too many long-lived tables live throughout <u>the entire</u> <u>lifespan</u> (<u>Max Duration</u>) of the database

## Tables are mostly thin

- On average, half of the tables (approx. 47%) are thin tables with less than 5 attributes.
- The tables with 5 to 10 attributes are approximately one third of the tables' population
- The large tables with more than 10 attributes are approximately 17% of the tables.

Pct of tables with num. of attributes ...

	<u>&lt;5</u>	<u>5-10</u>	<u>&gt;10</u>
atlas	10,23%	68,18%	21,59%
biosql	75,56%	24,44%	0,00%
coppermine	52,17%	30,43%	17,39%
ensembl	54,84%	38,06%	7,10%
mediawiki	61,97%	19,72%	18,31%
phpbb	40,00%	44,29%	15,71%
typo3	21,88%	31,25%	46,88%
opencart	57,20%	33,05%	9,75%
Average	46,73%	36,18%	17,09%

#### Schema size @ birth / duration

If you 're wide, you survive a.k.a (only the thin die young, <del>all</del> the wide ones seem to live forever)

### THE GAMMA PATTERN

## The Gamma **F** Pattern: "if you 're wide, you survive"

- The Gamma phenomenon:
  - tables with small schema sizes can have arbitrary durations, //small size does not determine duration
  - larger size tables last long
- Observations:
  - whenever a table exceeds the critical value of 10 attributes in its schema, its chances of surviving are high.
  - in most cases, the large tables are created early on and are not deleted afterwards.



















#### Exceptions

10

- Biosql: nobody exceeds 10 attributes
- Ensembl, mwiki: very few exceed 10 attributes, 3 of them died
- typo: has many late born survivors



### Stats on wide tables and their survival

			As pct o	ver #Tables	es As pct over the set of Wide Tables					
	#	# Wide		Wide of long		Early Born	of Long			
	Tables	tables	Wide	Wide duration		& Survivors	Duration			
coppermine	23	4	17%	17%	100%	100%	100%			
phpBB	70	11	16%	14%	<b>91</b> %	91%	91%			
opencart*	128	12	9%	7%	100%	75%	75%			
atlas	88	14	16%	11%	86%	71%	71%			
typo3	32	15	47%	13%	<b>87</b> %	33%	27%			
mwiki	71	6	8%	1%	50%	33%	17%			
ensembl	155	9	6%	0%	67%	56%	0%			
biosql	45	0	0%	0%	NA	NA	NA			

#### **Definitions**:

Wide schema: strictly above 10 attributes.

**The top band of durations** (the upper part of the Gamma shape): the upper 10% of the values in the y-axis.

Early born table: ts birth version is in the lowest 33% of versions;

Late-comers: born after the 77% of the number of versions.

# Whenever a table is wide, its chances of surviving are high

			As pct o	over #Tables	As pct over the set of Wide Tables					
	#	# Wide		Wide of long		Early Born	of Long			
	Tables	tables	Wide	duration	Survivors	& Survivors	Duration			
coppermine	23	4	17%	17%	100%	100%	100%			
phpBB	70	11	16%	14%	91%	91%	91%			
opencart*	128	12	9%	7%	100%	75%	75%			
atlas	88	14	16%	11%	86%	71%	71%			
typo3	32	15	47%	13%	87%	33%	27%			
mwiki	71	6	8%	1%	50%	33%	17%			
ensembl	155	9	6%	0%	67%	56%	0%			
biosql	45	0	0%	0%	NA	NA	NA			

Apart from mwiki and ensembl, all the rest of the data sets *confirm the hypothesis with a percentage higher than 85%*. The two exceptions are as high as 50% for their support to the hypothesis.

## Wide tables are frequently created early on and are not deleted afterwards

			As pct over #Tables		As pct over the set of Wide Tables					
	#	# Wide		Wide of long		Early Born	of Long			
	Tables	tables	Wide duration		Survivors	& Survivors	Duration			
coppermine	23	4	17%	17%	100%	100%	100%			
phpBB	70	11	16%	14%	<b>91</b> %	91%	91%			
opencart*	128	12	9%	7%	100%	75%	75%			
atlas	88	14	16%	11%	86%	71%	71%			
typo3	32	15	47%	13%	<b>87</b> %	33%	27%			
mwiki	71	6	8%	1%	50%	33%	17%			
ensembl	155	9	6%	0%	67%	56%	0%			
biosql	45	0	0%	0%	NA	NA	NA			

Early born, wide, survivor tables (as a percentage over the set of wide tables).

- in half the data sets the percentage is above 70%
- in two of them the percentage of these tables is one third of the wide tables.

## Whenever a table is wide, its duration frequently lies within the top-band of durations (upper part of Gamma)

			As pct	over #Tables	As pct of	over the set of Wid	de Tables
	#	# Wide		Wide of long		Early Born	of Long
	Tables	tables	Wide	duration	Survivors	& Survivors	Duration
coppermine	23	4	17%	17%	100%	100%	100%
phpBB	70	11	16%	14%	91%	91%	91%
opencart*	128	12	9%	7%	100%	75%	75%
atlas	88	14	16%	11%	86%	71%	71%
typo3	32	15	47%	13%	87%	33%	27%
mwiki	71	6	8%	1%	50%	33%	17%
ensembl	155	9	6%	0%	67%	56%	0%
biosql	45	0	0%	0%	NA	NA	NA

What is probability that a wide table belongs to the upper part of the Gamma?

- there is a very strong correlation between the two last columns: the Pearson correlation is 88% overall; 100% for the datasets with high pct of early born wide tables.

- Bipolarity on this pattern: half the cases support the pattern with support higher than 70%, whereas the rest of the cases clearly disprove it, with very low support values.

### Long-lived & wide => early born and survivor

			As pct o	As pct over #Tables Wide of long		r #Tables As pct over the set of Wide Tables					
	#	# Wide				Early Born	of Long				
	Tables	tables	Wide	duration	Survivors	& Survivors	Duration				
coppermine	23	4	17%	17%	100%	100%	100%				
phpBB	70	11	16%	14%	<b>91</b> %	91%	91%				
opencart*	128	12	9%	7%	100%	75%	75%				
atlas	88	14	16%	11%	86%	71%	71%				
typo3	32	15	47%	13%	<b>87</b> %	33%	27%				
mwiki	71	6	8%	1%	50%	33%	17%				
ensembl	155	9	6%	0%	67%	56%	0%				
biosql	45	0	0%	0%	NA	NA	NA				
						A.	Subset				

relationship

In all data sets, if a <u>wide</u> table has a <u>long duration</u> within the <u>upper part of the</u> <u>Gamma</u>, this deterministically (100% of all data sets) signifies that the table was also <u>early born</u> and <u>survivor</u>.

If a wide table is in the top of the Gamma line, it is deterministically an early born survivor.

Schema size and updates

### THE COMET PATTERN

## The Comet Pattern

"Comet " for change over schema size with:

- a large, dense, nucleus cluster close to the beginning of the axes, denoting small size and small amount of change,
- medium schema size tables typically demonstrating medium to large change
  - The tables with the largest amount of change are typically tables whose schema is on average one standard deviation above the mean
- wide tables with large schema sizes demonstrating small to medium (typically around the middle of the yaxis) amount of change.





















http://visual.merriam-webster.com/astronomy/celestial-bodies/comet.php



Comets have two tails: White one is made of comet dust particles. Blue one is made of electrically charged gas The coma is the cloud of comet dust particles surrounding the nucleus Nucleus is solid, icy heart of comet, inside the cloud of the coma

http://spaceplace.nasa.gov/comet-nucleus/en/

## Statistics of schema size at birth and sum of updates

			Schema size at birth					Sum of updates				
	#tables	max	mean (μ)	stdev (σ)	median	mode	ı	max	mean (μ)	stdev (σ)	median	mode
atlas	87 / 88	24	7.53	3.67	7	6		32	5.86	11.81	4	0
biosql	45	8	3.6	1.37	3	2		22	5.38	11.91	1	0
coppermine	23	25	6.52	5.35	4	2		18	3.3	7.98	1	0
ensembl	155	16	4.98	2.98	4	3		87	10.38	27.05	3	0
mwiki	71	20	4.79	3.64	3	3		43	6.92	16.03	3	0
ocart*	128	53	5.73	7.02	4	3		42	2.56	8.56	0	0
phpBB	70	98	9.39	14.63	5	3		97	6.33	22.17	0.5	0
typo3	32	30	12.69	9.26	8.5	4		61	7.53	20.89	1.5	0

/\* atlas: excluded table l1\_prescale\_set from the analysis (266 attributes; second largest value: 24); open cart: after version 22\*/

### Typically: ~70% of tables inside the box

		In the box		Out of th	the box	
	#tables	count	pct	count	pct	
atlas	88	62	70%	26	30%	
biosql	45	31	69%	14	31%	
coppermine	23	18	78%	5	22%	
ensembl	155	100	65%	55	35%	
mwiki	71	50	70%	21	30%	
ocart*	128	110	86%	18	14%	
phpBB	70	51	73%	19	27%	
typo3	32	16	50%	16	50%	

/\* atlas: excluded table l1\_prescale\_set from the analysis (266 attributes; second largest value: 24); open cart: after version 22\*/

Typically, around 70% of the tables of a database is found within the 10x10 box of *schemaSize@birth* x *sumOfUpdates* (10 excluded in both axes).

## Top changers tend to have medium schema sizes

Schema size @	birth.							
Statistics for			the ent	ire data set	the top changers			
	#tables	max	mean (µ)	stdev (σ)	μ+σ	avg sc. size for top 5%	sc. size of top 1	avg top 5% / max
atlas⁻⁻	87	24	7.53	3.67	11.20	9.60	6	0.40
biosql	45	8	3.60	1.37	4.97	5.00	5	0.63
coppermine	23	25	6.52	5.35	11.87	12.50	14	0.50
ensembl	155	16	4.98	2.98	7.97	7.13	10	0.45
mwiki	71	20	4.79	3.64	8.43	8.25	13	0.41
ocart*	128	53	5.73	7.02	12.74	17.43	39	0.33
phpBB	70	98	9.39	14.63	24.02	48.00	98	0.49
typo3	32	30	12.69	9.26	21.95	19.50	19	0.65
Pearson with								
avg top 5%		0.96	0.58	0.97	0.87		0.97	

/\* atlas: excluded table l1\_prescale\_set from the analysis (266 attributes; second largest value: 24); open cart: after version 22\*/

For every dataset: we selected the top 5% of tables in terms of this sum of updates and we averaged the schema size at birth of these top 5% tables.

## Top changers tend to have medium schema sizes

Schema size @	birth.								
Statistics for		the entire data set				the top changers			
	#tables	max	mean (µ)	stdev (σ)	μ+σ	avg sc. size for top 5%	sc. size of top 1	avg top 5% / max	
atlas⁻⁻	87	24	7.53	3.67	11.20	9.60	6	0.40	
biosql	45	8	3.60	1.37	4.97	5.00	5	0.63	
coppermine	23	25	6.52	5.35	11.87	12.50	14	0.50	
ensembl	155	16	4.98	2.98	7.97	7.13	10	0.45	
mwiki	71	20	4.79	3.64	8.43	8.25	13	0.41	
ocart*	128	53	5.73	7.02	12.74	17.43	39	0.33	
phpBB	70	98	9.39	14.63	24.02	48.00	98	0.49	
typo3	32	30	12.69	9.26	21.95	19.50	19	0.65	
Pearson with									
avg top 5%		0.96	0.58	0.97	<u>0.87</u>		0.97		

/\* atlas: excluded table l1\_prescale\_set from the analysis (266 attributes; second largest value: 24); open cart: after version 22\*/

The average schema size for the top 5% of tables in terms of their update behavior is close to one standard deviation up from the average value of the schema size at birth(i.e., very close to \$mu\$+\$sigma\$). //except phpBB

## Top changers tend to have medium schema sizes

Schema size @	birth.							
Statistics for			the ent	ire data set	the top changers			
	#tables	max	mean (μ)	stdev (σ)	μ+σ	avg sc. size for top 5%	sc. size of top 1	avg top 5% / max
atlas⁻⁻	87	24	7.53	3.67	11.20	9.60	6	0.40
biosql	45	8	3.60	1.37	4.97	5.00	5	0.63
coppermine	23	25	6.52	5.35	11.87	12.50	14	0.50
ensembl	155	16	4.98	2.98	7.97	7.13	10	0.45
mwiki	71	20	4.79	3.64	8.43	8.25	13	0.41
ocart*	128	53	5.73	7.02	12.74	17.43	39	0.33
phpBB	70	98	9.39	14.63	24.02	48.00	98	0.49
typo3	32	30	12.69	9.26	21.95	19.50	19	0.65
Pearson with								
avg top 5%		0.96	0.58	0.97	0.87		0.97	

/\* atlas: excluded table l1\_prescale\_set from the analysis (266 attributes; second largest value: 24); open cart: after version 22\*/

In 5 out of 8 cases, the average schema size of top-changers within 0.4 and 0.5 of the maximum value (practically the middle of the domain) and never above 0.65 of it.
Pearson: the maximum value, the standard deviation of the entire data set and the average of the top changers are very strongly correlated.
#### Wide tables have a medium number of updates

Total amt. of up Statistics for	the entire data set					the top 5% with respect to schema size at birth (top wide)				
	#tables	max	mean (μ)	stdev (σ)	μ+σ	max/2	avg upd. of top 5%	upd. of top 1	avg of top 5% / max	Top up. in wide?
atlas	88	32	5.86	11.81	11. <mark>81</mark>	16.0	12.60	20	0.39	Ν
biosql	45	22	5.38	11.91	11.91	11.0	8.00	0	0.36	Ν
coppermine	23	18	3.30	7.98	7.98	9.0	13.50	9	0.75	Y
ensembl	155	87	10.38	27.05	27.05	43.5	28.22	0	0.32	Ν
mwiki	71	43	6.92	16.03	16.03	21.5	17.75	19	0.41	Y
ocart*	128	42	2.56	8.56	8.561	21.0	14.55	2	0.35	Y
phpBB	70	97	6.33	22.17	22.17	48.5	43.00	97	0.44	Y!
typo3	32	61	7.53	20.89	20.89	30.5	2.00	1	0.03	Ν
Pearson with avg top 5%			0.27	0.59	0.50	0.74		0.79		

For each data set, we took the top 5% in terms of schema size at birth (**top wide**) and contrasted their update behavior wrt the update behavior of the entire data set. Typically, the avg. number of updates of the top wide tables is close to the 38% of the domain of values for the sum of updates (i.e., the middle of the y-axis of the comet figure, measuring the sum of updates for each table).

This is mainly due to the (very) large standard deviation (twice the mean), rather than the -- typically low -- mean value (due to the large part of the population living quiet lives).

#### **INVERSE GAMMA**

# The inverse Gamma pattern

- The correlation of change and duration is as follows:
  - small durations come necessarily with small change,
  - large durations come with all kinds of change activity and
  - medium sized durations come mostly with small change activity (Inverse Gamma).























### Skyline & Avg for Inverse Gamma

#### THE EMPTY TRIANGLE PATTERN

#### Quiet tables rule, esp. for mature db's

Table distribution (pct of tables) wrt their activity class

			DIEI			SURVIVED				Aggregate per update type		
	#tables	No change	Quiet	Active	Total	No change	Quiet	Active	Total	No change	Quiet	Active
atlas	88	8%	7%	2%	17%	13%	42%	28%	83%	20%	49%	31%
biosql	45	20%	13%	4%	38%	16%	16%	31%	62%	36%	29%	36%
phpbb	70	0%	3%	4%	7%	50%	31%	11%	93%	50%	34%	16%
typo3	32	16%	6%	6%	28%	22%	34%	16%	72%	38%	41%	22%
coppermine	23	4%	0%	0%	4%	30%	57%	9%	96%	35%	57%	<b>9%</b>
ensembl	155	24%	20%	8%	52%	6%	35%	7%	48%	30%	55%	15%
mwiki	71	14%	13%	3%	30%	3%	63%	4%	70%	17%	76%	7%
opencart*	128	9%	2%	0%	11%	42%	44%	3%	89%	51%	46%	3%

#### Non-survivors

- Sudden deaths mostly
- Quiet come ~ close
- Too few active

#### <u>Survivors</u>

- Quiet tables rule
- Rigid and active then
- Active mostly in "new" db's

Mature DB's: the pct of active tables drops significantly



#### mwiki: duration / birth



## Survive long enough & you 're probably safe

It is quite rare to see tables being removed at old age Typically, the area of high duration is overwhelmingly inhabited by survivors (although each data set comes with a few such cases)!





#### Die young and suddenly

[Early life of the db] There is a very large concentration of the deleted tables in a small range of newly born, quickly removed, with few or no updates, resulting in very low numbers of removed tables with medium or long durations.

[Mature db] After the early stages of the databases, we see the birth of tables who eventually get deleted, but they mostly come with very small durations and sudden deaths.









▲ Sudden Death

O Quiet, Dead

Active, Dead

Quiet Survivor

Active Surviror

Sudden Death

O Quiet, Dead

Active, Dead

Quiet Survivor

Active Surviror

20

500

▲ Rigid

Δ

40

▲ Rigid





## Top changers: early born, survivors, often with long durations, and often all the above

	atlas	biosql	coppermine	ensembl	mwiki	ocart*	phpBB	typo3
Tables	88	45	23	155	71	128	70	32
Active	27	16	2	23	5	4	11	7
active tables(%)	31%	36%	9%	15%	7%	3%	16%	22%
As percentages over active								
Born early	96%	81%	100%	78%	80%	75%	82%	86%
Survivors	93%	88%	100%	48%	60%	100%	73%	71%
Long duration	85%	69%	100%	22%	40%	75%	55%	57%
Born early, survive, live long	85%	69%	100%	22%	40%	75%	55%	57%

- In all data sets, active tables are born early with percentages that exceed 75%
- With the exceptions of two data sets, they survive with percentage higher than 70%.
- The probability of having a long duration is higher than 50% in 6 out of 8 data sets.
- Interestingly, the two last lines are exactly the same sets of tables in all data sets!
  - An active table with long duration has been born early and survived with prob. 100%
  - An active, survivor table that has a long duration has been born early with prob. 100%

# Dead are: quiet, early born, short lived, and quite often all three of them

	atlas	biosql	coppermine	ensembl	mwiki	ocart*	phpBB	typo3
tables	88	45	23	155	71	128	70	32
dead	15	17	1	80	21	14	5	9
dead tables(%)	17%	38%	4%	52%	30%	11%	7%	28%
As percentages over # dead								
· · ·								
Few updates	87%	88%	100%	85%	90%	100%	40%	78%
Early born	80%	82%	100%	70%	62%	71%	100%	78%
Short-lived	80%	76%	0%	89%	90%	100%	0%	22%
Few upd's, early born,	600/	E 00/	09/	E10/	420/	710/	09/	00/
short duration	60%	59%	0%	51%	43%	71%	0%	0%
Do tables die of old age?								
long durations	48	14	18	13	23	86	57	12
long durations, dead	0	0	0	0	1	0	0	0
Dead among long-lived (%)	0%	0%	0%	0%	4%	0%	0%	0%





## Longevity and update activity correlate !!

- Remember: top changers are defined as such wrt ATU (AvgTrxnUpdate), not wrt sum(changes)
- Still, they dominate the sum(updates) too! (see top of inverse  $\Gamma$ )
- See also upper right
   blue part of diagonal:
   too many of them
   are born early and
   survive => live long!



#### All in one

- Early stages of the database life are more
   "active" in terms of births, deaths and updates, and have higher chances of producing deleted tables.
- After the first major restructuring, the database continues to grow; however, we see much less removals, and maintenance activity becomes more concentrated and focused.

... How do survivor tables differ from the dead ones (esp., wrt activity & duration)?

http://www.cs.uoi.gr/~pvassil/publications/2017 CAiSE Electrolysis

Panos Vassiliadis, Apostolos Zarras. 29th International Conference on Advanced Information Systems Engineering, (CAiSE 2017), 12-16 June 2017, Essen, Germany.

#### SURVIVAL IN SCHEMA EVOLUTION: PUTTING THE LIVES OF SURVIVOR AND DEAD TABLES IN COUNTERPOINT

# ELECTROLYSIS PATTERN FOR TABLE





#### Duration is related to the Life & Death Class of the tables!



- (a) Survival: DEAD vs SURVIVORS
- (b) Activity: <u>Rigid</u> (no change) vs <u>Active</u> (change rate > 10%) vs <u>Quiet</u> (all in between)
- (c) Life And Death (LAD) class: Survival x Activity



6

Year Range

pct for class

pct for class

Year Range

4

127

0 4

### The electrolysis pattern



- Dead tables demonstrate much shorter lifetimes than survivor ones,
- can be located at short or medium durations, and practically never at high durations.
- With few exceptions, the less active dead tables are, the higher the chance to reach shorter durations.
- Survivors expose the inverse behavior, i.e., mostly located at medium or high durations.
- The more active survivors are, the stronger they are attracted towards high durations, with a significant such inclination for the few active ones that cluster in very high durations.

#### The electrolysis pattern: survivors



- The extreme clustering of active survivors to high durations
- The wider spread of (quite numerous) quiet survivors to a large span of durations with long trails of points
- The clustering of rigid survivors, albeit not just to one, but to all kinds of durations (frequently, not as high as quiet and active survivors)

#### The electrolysis pattern: dead



- The total absence of dead tables from high durations
- The clustering of rigid dead at low durations,
- the spread of quiet dead tables to low or medium durations, and
- the occasional presence of the few active dead, that are found also at low or medium durations, but in a clustered way

## Electrolysis as a heatmap showing the extreme bias between dead and survivor tables



- For each *LifeAndDeath* value, and for each duration range of 5% of the database lifetime, we computed the percentage of tables (over the total of the data set) whose duration falls within this range.
- We removed cells that corresponded to only one data set

The resulting heatmap shows the polarization in colors: brighter color signifies higher percentage of the population

### Gravitation to Rigidity



- Although the majority of survivor tables are in the quiet class, we can quite emphatically say that it is the absence of evolution that dominates!
  - Survivors vastly outnumber removed tables.
  - Similarly, rigid tables outnumber the active ones, both in the survival and, in particular, in the dead class.
  - Schema size is rarely resized, and only in survivors (not in the paper).
  - Active tables are few and do not seem to be born in other but early phases of the database lifetime.
- Evidently, not only survival is also stronger than removal, but **rigidity is also stronger a force than variability** and the combination of the two forces further lowers the amount of change in the life of a database schema.

### Electrolysis

- Yes, we can indeed find **patterns in the lives of tables**, during schema evolution!
- <u>Survivors</u>, mostly <u>long-lived</u> (esp. active ones) and <u>quietly active</u> are <u>radically different</u> than <u>dead</u> tables, being mostly <u>short-lived</u> and <u>rigid</u>!
- Gravitation to rigidity rules: we see more absence than presence of schema evolution!



#### Also studied [not part of the paper]: year of birth, schema size, schema resizing

... How do foreign keys evolve?

http://www.cs.uoi.gr/~pvassil/publications/2017\_ER

Panos Vassiliadis, Michail-Romanos Kolozoff\*, Maria Zerva, Apostolos V. Zarras. 36th International Conference on Conceptual Modeling **(ER 2017)**, Nov. 6th-9th, 2017, Valencia Spain

#### SCHEMA EVOLUTION AND FOREIGN KEYS: BIRTH, EVICTION, CHANGE AND ABSENCE

#### **Research Questions**

- How do foreign keys evolve over time?
  - Do tables and foreign keys <u>evolve in sync</u>?
  - <u>When</u> & <u>How</u> do foreign keys <u>germinate & die</u>?
- ... as we will see, these questions led to unexpected results and more insights on how developers deal with foreign keys...



### Evolution of Tables & FK's

- Tables grow in all cases
  (known from previous
  research) with periods
  of slow growth,
  calmness, spikes of
  extension, and
  occasional cleanups
- Foreign Keys are treated with different mentalities. 3 families:
  - Scientific
  - Comp. Toolkits
  - CMS's

#### Evolution of Tables & FK's: Scientific projects



- Tables and FKS grow in synch, in both cases
- Growth comes with expansion periods, shrinkage actions, and periods of calmness in terms of both tables and foreign keys.

#### Evolution of Tables & FK's: Computational Resource Toolkits



- Tables and FKS grow little and slowly; for Castor, not exactly in sync
- Castor: observe how scarce FK's are (too few tables come with FK's, see vertical axis)

### Evolution of Tables & FK's: Content Management Systems (CMS's)



- FK scarcity: really big at Slashcode, moderate at Zabbix
- Slashcode started <u>without</u> foreign keys at all; 1<sup>st</sup> set of FK's in v. 74.
   Zabbix seems to show a certain degree of syncronized growth
- Yet, ... both CMS's end up with no FK's!! -> see next

## What an unpleasant surprise: developers can resort in full removal of foreign keys!



- Slashcode: there is a clear phase of progressive removal
- Zabbix: abrupt removal of almost the entire set of foreign keys in a single transition. We have <u>no knowledge on why</u> this happened, & it is <u>unexpected</u> based on how FK's had been treated till then... 140

### How do FK's germinate and die?

- We classified FK's births and deaths in 4 categories
- Births
  - Born with table: when either the source or the target table is born along with the foreign key,
  - Explicit addition: when a foreign key is added to two existing tables.
- Deletions
  - Died with table: when either the source or the target table is removed along with the foreign key,
  - Explicit deletion: when neither of the source or target tables gets deleted and only the foreign key is removed.

#### Stats on FK Change

			Atlas	Biosql	Egee	Castor	Slashcode	Zabbix
	Diachronic Graph	TablesDG	88	45	12	91	126	58
	Diachronic Graph	FK'sDG	88	79	6	13	47	38
	Start/End	FKs@start	61	17	3	6	0	10
	Start/Enu	FKs@end	65	52	5	10	0	2
#FKs_added		Total Born w/	41	81	4	8	77	28
	in absolute numbers	table	37	71	3	2	21	24
		Explicit addition	4	10	1	6	56	4
	as pct	(%)Born w/ table (%)Explicit	90%	88%	75%	25%	27%	86%
		addition	10%	12%	25%	75%	73%	14%
		Total Died w/	37	46	2	4	77	36
	in absolute numbers	table	25	42	2	2	16	8
#FKs_removed		Explicit deletion	12	4	0	2	61	28
	as pct	(%)Died w/ table (%)Explicit	68%	91%	100%	50%	21%	22%
		deletion	32%	9%	0%	50%	79%	<sup>142</sup> 78%

#### Stats on FK Change

			Atlas	Biosql	Egee	Castor Slashcode Zabbix
	Diachronic Graph	TablesDG	88	45	12	
	Diachtonic Graph	FK'sDG	88	79	6	Atlas, Biosql and
	Start/End	FKs@start	61	17	3	Egee (less) deal
	StartyEnd	FKs@end	65	52	5	with <b>FK's as</b>
		Total	41	81	4	regular part of
#FKs_added	in absolute numbers as pct	Born w/ table Explicit	37	71	3	the schema
		addition	4	10	1	
		(%)Born w/ table (%)Explicit addition	90% 10%	<mark>88%</mark> 12%	<mark>75%</mark> 25%	FK's are, to a large extent - Born with
	in absolute	Total Died w/ table	37 25	46 42	2 2	tables - Removed with
	numbers	Explicit	25	42	2	tables
#FKs_removed		deletion	12	4	0	
	as pct	(%)Died w/ table (%)Explicit	68%	91%	100%	
		deletion	32%	9%	0%	

#### Stats on FK Change

			Atlas	Biosql	Egee	Castor	Slashcode	Zabbix
	Diachronic Graph	TablesDG	Center			91	126	58
		FK'sDG	Castor	& Slas	ncode	13	47	38
	Start/End	FKs@start	(both	with a r	really	6	0	10
	<b>,</b>	FKs@end	small I	- <u>minorit</u>	<u>y of</u>	10	0	2
#FKs_added		Total Born w/ table Explicit addition	<u>FK's</u> ) deal with <b>FK's</b>			8	77	28
	in absolute numbers		as an a	ad-hoc	add	2	21	24
			on: FK	's are n	nostly	6	56	4
	Г	(%)Born w/	explicitly added/					
	as pct	table (%)Explicit addition	remov	red		25%	27%	86%
						75%	73%	14%
		Total Died w/ table Explicit deletion	Zabbix has a mixed style: explicit del.			4	77	36
	in absolute numbers		•			2	16	8
#FKs_removed	-		and add. w. tables (& a sudden style			2	61	28
	as pct	(%)Died w/ table (%)Explicit deletion	<u>chang</u>	<u>e</u> )		50%	21%	22%
						50%	79%	78%
Families of developer profiles wrt the treatment of Foreign Keys

- Integral part of schema: fairly large pct of tables involved in FKs, grow in sync with tables, germinate and die with them
- Disposable Add-on: small pct of tables involved in FK's, explicit additions and deletions, easy to remove them (in some cases, entirely!)
- **Mixed**: can be with a change of style





### Percentage of transitions with FK change

	Total # transitions	Total # transitions with FK change	Pct. of transitions with FK change
Atlas	85	25	29%
BioSQL	46	19	41%
Egee	16	3	19%
Castor	191	6	3%
Slashcode	398	34	9%
Zabbix	159	22	14%

Common theme in all the data sets: the **consistent scarcity of FK changes** 

- Scientific data sets: short active period + treatment of FK's as an integral part of the schema (births and deaths of tables and FK's in sync) => high pct of transitions with FK change
- The rest: FK b&d are rare and explicit (w/o mass removals, would be even less)

Characteristics of the heartbeat of schemata wrt Foreign Keys

- Scarcity of FK change: expectedly very few transitions come with FK change, except for idiosyncratic cases
- Low volume: typically 1 FK change at a time, except for mass add/del
- Birth & deaths are proportionally spread in time
- Occasional do-undo and restructuring due to table renames



### Slashcode: the disappearing FK's



"Commented-out foreign keys are ones which currently cannot be used because they refer to a primary key which is NOT NULL AUTO INCREMENT and the child's key either has a default value which would be invalid for an auto increment field, typically NOT NULL DEFAULT '0'.

Or, in some cases, the primary key is e.g. VARCHAR(20) NOT NULL and the child's key will be VARCHAR(20). The possibility of NULLs negates the ability to add a foreign key. <= That's my current theory, but it doesn't explain why discussions.topic SMALLINT UNSIGNED NOT NULL DEFAULT '0' is able to be foreign-keyed to topics.tid SMALLINT UNSIGNED NOT NULL AUTO INCREMENT"

"Stories is now InnoDB and these other tables are still MyISAM, so no foreign keys between them."

"This doesn't work, makes createStory die. These don't work, should check why..."

"This doesn't work, since in the install pollquestions is populated before users, alphabetically"

"This doesn't work, since discussion may be 0."

1<sup>st</sup> massive foreign key removal (rev 1.120), 22 FK's deleted.

2<sup>nd</sup> massive deletion (rev 1.151), 10 FK's deleted

3<sup>rd</sup> deletion (rev 1.174), 3 FK's deleted

4<sup>th</sup> deletion (rev 1.189) 1 FK deleted

5<sup>th</sup> deletion (rev 1.201) 1 FK<sup>5</sup>deleted

# Slashcode: what did the comments say?

- The main problem seems to be the difficulty of developers with the tuning and handling of both foreign and primary keys.
- Sometimes <u>difficulties are hard</u> -- e.g., different storage engines, typically due to performance reasons
- Some difficulties are complicated <u>due to technicalities</u> like autonumbering
- Sometimes <u>fixes could be found with some effort (e.g.,</u> changing the order of table population, or using numeric data types for primary keys, or inserting some "goalkeeper" values at FK target table)

## Scarcity of Foreign keys

A 2013 collection of schema histories, lists 21 data sets,
 -- some have more than one target DBMS variants.

<pre>\$ cd RESEARCH/Github/EvolutionDatasets</pre>					
\$ ls -d * */*					
CERN	CMS's/Coppermine	CMS's/XOOPS	Med		
CERN/Atlas	CMS's/DekiWiki	CMS's/Zabbix	Med/Ensembl		
CERN/CASTOR	CMS's/Joomla 1.5	CMS's/e107	Med/biosql		
CERN/DQ2	CMS's/NucleusCMS	CMS's/opencart	README.md		
CERN/DRAC	CMS's/SlashCode	CMS's/phpBB			
CERN/EGEE	CMS's/TikiWiki	CMS's/phpwiki			
CMS's	CMS's/Typo3	CMS's/wikimedia			

How many data sets contain foreign keys?

Try this (also backed by manual sampling):

grep -rl "FOREIGN" . >> ALL-FKs-by-grep.ascii
awk '{split(\$0,a,"/"); print a[2],a[3]}' ALL-FKs-by-grep.ascii
uniq

## Scarcity of Foreign keys

### - How many data sets, out of the 21, contain foreign keys?

CERN Atlas CERN CASTOR CERN EGEE CMS's SlashC CMS's Zabbix Med biosql

CERN DQ2 CERN DIRAC Med Ensembl

### The 6 data sets reported here

**DQ2** (only in the mySQL, not in the Oracle version): FK's in 19 versions out of the 55. Starts with 2 FK's and ends with 1.

DIRAC (not in the production folder, only at python+mysql).
9 tables at first version, 15 tables at last version
Starts with 10 FK's, ends with 8

**Ensembl**: not able to link FK DDL files to table evolution, yet

 - 9 out of the 21 data sets do (including 3 that are really small for harnessing valuable results, spec., Egee, DQ2, DIRAC) you're not welcome here...

#sorrynotsorry



# Foreign Key Evolution comes with different treatments:

- Sometimes, **FK's are treated as an integral part of the system**, and they are born and evicted along with table birth and eviction.
- Other times, **FK's are treated as a disposable add-on**: only a small subset of the tables involved in FK's; birth and eviction of FK's rarely performed in synch with their tables. If technical difficulties arise, it is possible to witness the **complete removal of FK's** from the schema.
- Another sign of concern is that in all the CMS' we collected, FK's are too scarce
- More results in the paper: **stats**, **threats to validity**, and, the treatment of the **evolving schema as an evolving graph**

#### **Roadmap**

- Evolution of views
- Data warehouse Evolution
- A case study (if time)
- Impact assessment in ecosystems
- Empirical studies concerning database evolution
- Open Issues and discussions

... and data intensive ecosystems...

## IMPACT ASSESSMENT

## Data intensive ecosystems

- Ecosystems of applications, built on top of one or more databases and strongly dependent upon them
- Like all software systems, they too change...



## Evolving data-intensive ecosystem



## Evolving data-intensive ecosystem



The impact can be syntactical (causing crashes), semantic (causing info loss or inconsistencies) and related to the performance

## The impact of changes & a wish-list

- Syntactic: scripts & reports simply crash
- Semantic: views and applications can become inconsistent or information losing
- Performance: can vary a lot

We would like: evolution predictability i.e., control of what will be affected before changes happen

- Learn what changes & how
- Find ways to quarantine effects



## The **Hecataeus** tool & method. Here: a first map of Drupal



#### http://www.cs.uoi.gr/~pvassil/projects/hecataeus/

## What happens if I modify table search\_index? Who are the neighbors?



## What happens if I modify table search\_index? Who are the neighbors?



Tooltips with info on the script & query

## In the file structure too...



## How to handle evolution?



- Architecture Graphs: graph with the data flow between modules (i.e., relations, views or queries) at the detailed (attribute) level; module internals are also modeled as subgraphs of the Architecture Graph
- Policies, that annotate a module with a reaction for each possible event that it can withstand, in one of two possible modes:
  - (a) block, to veto the event and demand that the module retains its previous structure and semantics, or,
  - (b) propagate, to allow the event and adapt the module to a new internal structure.
- Given a potential change in the ecosystem
  - we identify which parts of the ecosystem are affected via a "change propagation" algorithm
  - we rewrite the ecosystem to reflect the new version in the parts that are affected and do not veto the change via a rewriting algorithm
    - Within this task, we resolve conflicts (different modules dictate conflicting reactions) via a conflict resolution algorithm

## University E/S Architecture Graph



## Architecture Graph



## Policies to predetermine reactions



Policies to predetermine the modules' reaction to a hypothetical event?

RELATION.OUT.SELF: on ADD\_ATTRIBUTE then PROPAGATE; RELATION.OUT.SELF: on DELETE\_SELF then PROPAGATE; RELATION.OUT.SELF: on RENAME\_SELF then PROPAGATE; RELATION.OUT.ATTRIBUTES: on DELETE\_SELF then PROPAGATE; RELATION.OUT.ATTRIBUTES: on RENAME SELF then PROPAGATE;

VIEW.OUT.SELF: on ADD\_ATTRIBUTE then PROPAGATE; VIEW.OUT.SELF: on ADD\_ATTRIBUTE\_PROVIDER then PROPAGATE; VIEW.OUT.SELF: on DELETE\_SELF then PROPAGATE; VIEW.OUT.SELF: on RENAME\_SELF then PROPAGATE; VIEW.OUT.ATTRIBUTES: on DELETE\_SELF then PROPAGATE; VIEW.OUT.ATTRIBUTES: on DELETE\_PROVIDER then PROPAGATE; VIEW.OUT.ATTRIBUTES: on DELETE\_PROVIDER then PROPAGATE; VIEW.OUT.ATTRIBUTES: on RENAME\_PROVIDER then PROPAGATE; VIEW.IN.SELF: on DELETE\_PROVIDER then PROPAGATE; VIEW.IN.SELF: on RENAME\_PROVIDER then PROPAGATE; VIEW.IN.SELF: on ADD\_ATTRIBUTE\_PROVIDER then PROPAGATE; VIEW.IN.ATTRIBUTES: on DELETE\_PROVIDER then PROPAGATE; VIEW.IN.ATTRIBUTES: on DELETE\_PROVIDER then PROPAGATE; VIEW.IN.SELF: on ADD\_ATTRIBUTE\_PROVIDER then PROPAGATE; VIEW.IN.ATTRIBUTES: on RENAME\_PROVIDER THEN PROPAGATE;

QUERY.OUT.SELF: on ADD\_ATTRIBUTE then PROPAGATE; QUERY.OUT.SELF: on ADD\_ATTRIBUTE\_PROVIDER then PROPAGATE; QUERY.OUT.SELF: on DELETE\_SELF then PROPAGATE; QUERY.OUT.ATTRIBUTES: on DELETE\_SELF then PROPAGATE; QUERY.OUT.ATTRIBUTES: on DELETE\_SELF then PROPAGATE; QUERY.OUT.ATTRIBUTES: on DELETE\_PROVIDER then PROPAGATE; QUERY.OUT.ATTRIBUTES: on DELETE\_PROVIDER then PROPAGATE; QUERY.OUT.ATTRIBUTES: on RENAME\_PROVIDER then PROPAGATE; QUERY.IN.SELF: on DELETE\_PROVIDER then PROPAGATE; QUERY.IN.SELF: on ADD\_ATTRIBUTE then PROPAGATE; QUERY.IN.SELF: on ADD\_ATTRIBUTE\_PROVIDER then PROPAGATE; QUERY.IN.ATTRIBUTES: on DELETE\_PROVIDER theN PROPAGATE; QUERY.IN.ATTRIBUTES: ON ADLATTRIBUTE\_PROVIDER theN PROPAGATE; QUERY.IN.ATTRIBUTES: ON ALTER\_SEMANTICS theN PROPAGATE;

## How to handle evolution?



## Internals of impact assess. & rewriting

- 1. Impact assessment. Given a potential event, a status determination algorithm makes sure that the nodes of the ecosystem are assigned a status concerning (a) whether they are affected by the event or not and (b) what their reaction to the event is (block or propagate).
- 2. Conflict resolution and calculation of variants. Algorithm that checks the affected parts of the graph in order to highlight affected nodes with whether they will adapt to a new version or retain both their old and new variants.
- **3. Module Rewriting**. Our algorithm visits affected modules sequentially and performs the appropriate restructuring of nodes and edges.

## Impact assessment & rewriting





## Conflicts: what they are and how to handle them (more than flooding)





- View0 initiates a change
- View1 and View 2 accept the change
- Query2 rejects the change
- Query1 accepts the change

- The path to Query2 is left intact, so that it retains it semantics
- View1 and Query1 are adapted
- View0 and View2 are adapted too, however, we need two version for each: one to serve Query2 and another to serve View1 and Query1

## Played an impact analysis scenario: delete attr. 'word' from search\_index

