

FINDING AND FIXING THE ARCHITECTURAL ROOTS OF BUGGINESS

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YOUR TYPICAL SOFTWARE PROJECT



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The boat is leaking but you keep paddling!

Why?

- 1. The illusion of progress.**
- 2. The lack of measurements.**

ARCHITECTURAL FLAWS



ARCHITECTURE MODELS

Do they help us understand an architecture?

Do they help us analyze an architecture?

Do they help us determine what is wrong with an architecture?





Mostly

No.

BRIDGING THE GAP BETWEEN ARCHITECTURE AND QUALITY

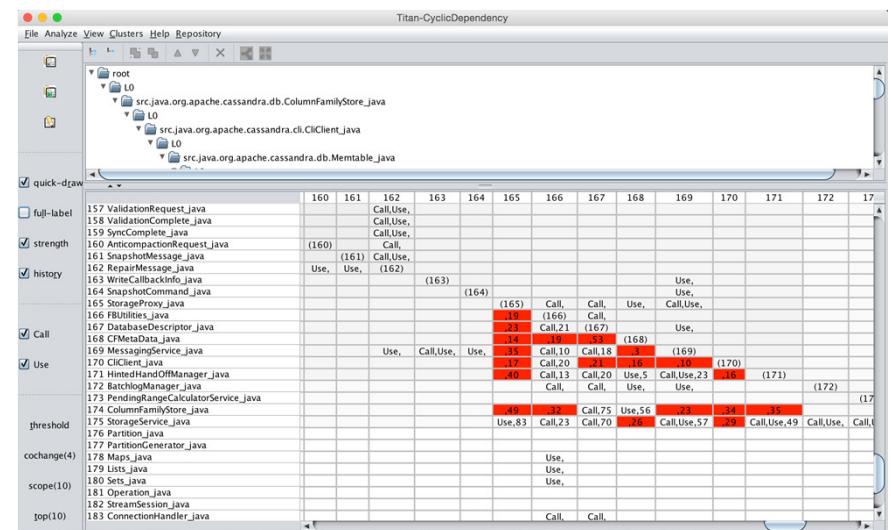
To bridge this gap we have created a design representation and associated tool:

Representation: Design Rule Space (DRSpace)

- Formed by selecting structural and/or evolutionary dependency relationships
- Clustered using the Design Rule Hierarchy algorithm

Tool: Titan

- Manipulates and visualizes DRSpaces
- Helps to *discover* architecture flaws



BRIDGING THE GAP BETWEEN ARCHITECTURE AND QUALITY

Using Titan we can find architecture design flaws:

- cyclic class dependencies
- cyclic package dependencies
- improper inheritance
- modularity violations
- unstable interfaces

Identifying such flaws allows us to:

- assess technical debt and its economic impact
- predict the economic impact of repair strategies

How does this work?



BACKGROUND CONCEPTS

Design Rule (DR)

- Design decisions that decouple the rest of the system into modules (typically a major interface or abstraction)

Design Structure Matrix (DSM)

- Square matrix; rows and columns are classes (or files)
- Cell(i,j) shows a dependency relationship between the class in the ith row and the class in the jth column

Design Rule Hierarchy (DRH)

- A DSM clustering algorithm
- Lower layers only depend on higher layers
- Modules in the same layers are independent of each other

DSM IN DRH STRUCTURE

Factory
pattern in a
simple maze
game

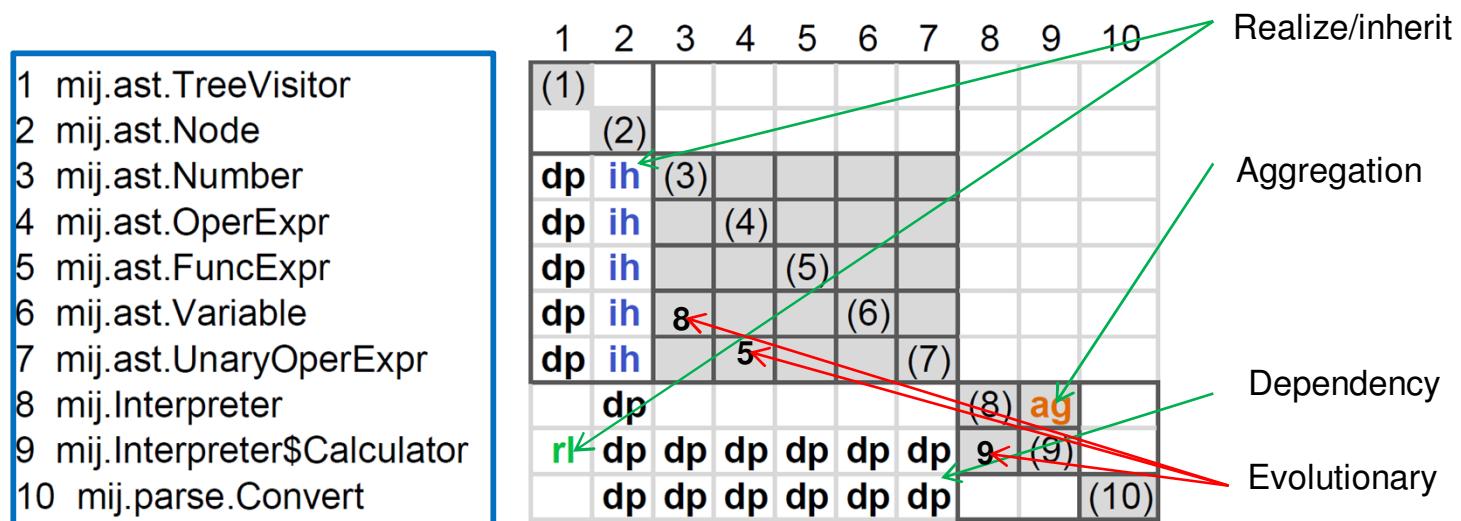
	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
MapSite	.														
Room	x	.													
Wall	x	.													
Maze		x	.												
Door	x	x	.												
MazeFactory															
BlackBlueDoor	x														
GreenRoom	x														
BlueWall	x														
BlueMazeFactory		x	x	x	x	x	x	x	.						
BrownDoor		x													
RedRoom	x														
RedWall	x														
RedMazeFactory		x	x	x											
SimpleMazeGame	x	x	x	x				x		x		x	.		

What is a DRSpace?



DESIGN RULE SPACE

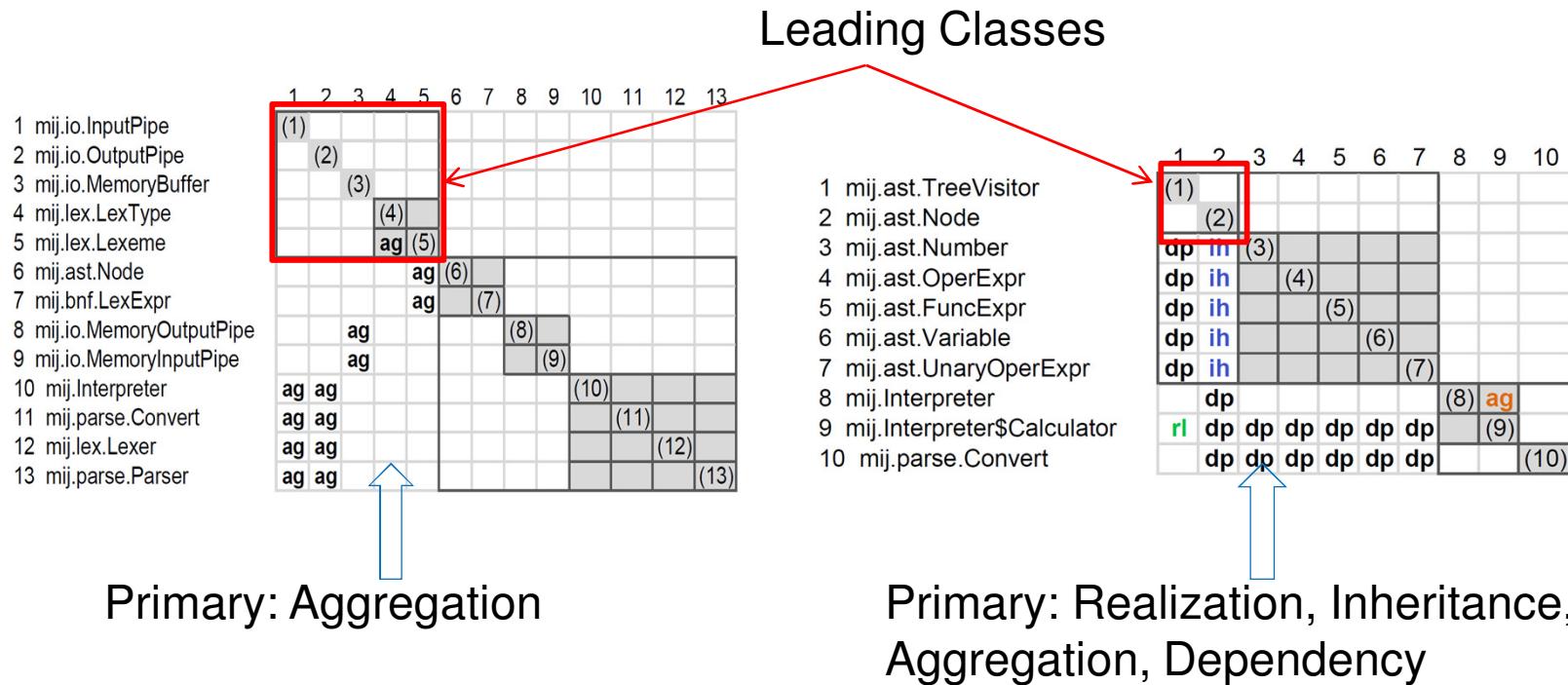
- A set of classes
 - A set of dependencies of different types revealing:
 - Structural relationships
 - Evolutionary relationships



DESIGN RULE SPACE

A DRSpace is clustered using the DRH algorithm, based on selected *primary dependency types*.

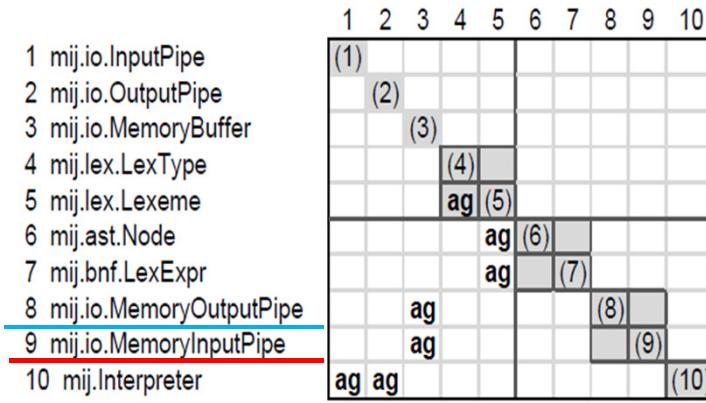
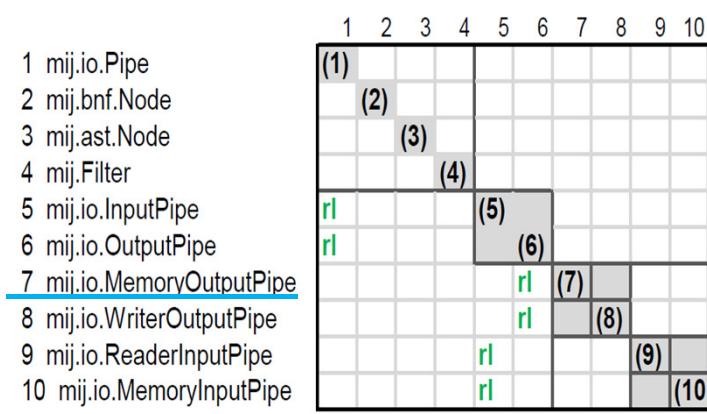
The leading classes are the *design rules*.



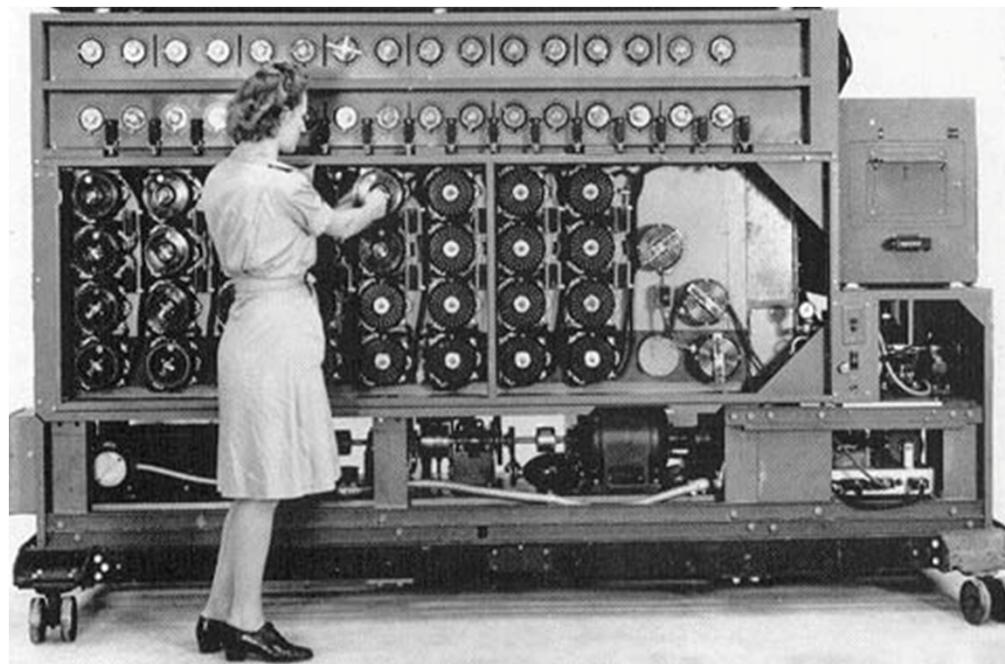
ARCHITECTURE AS A SET OF DRSPACES

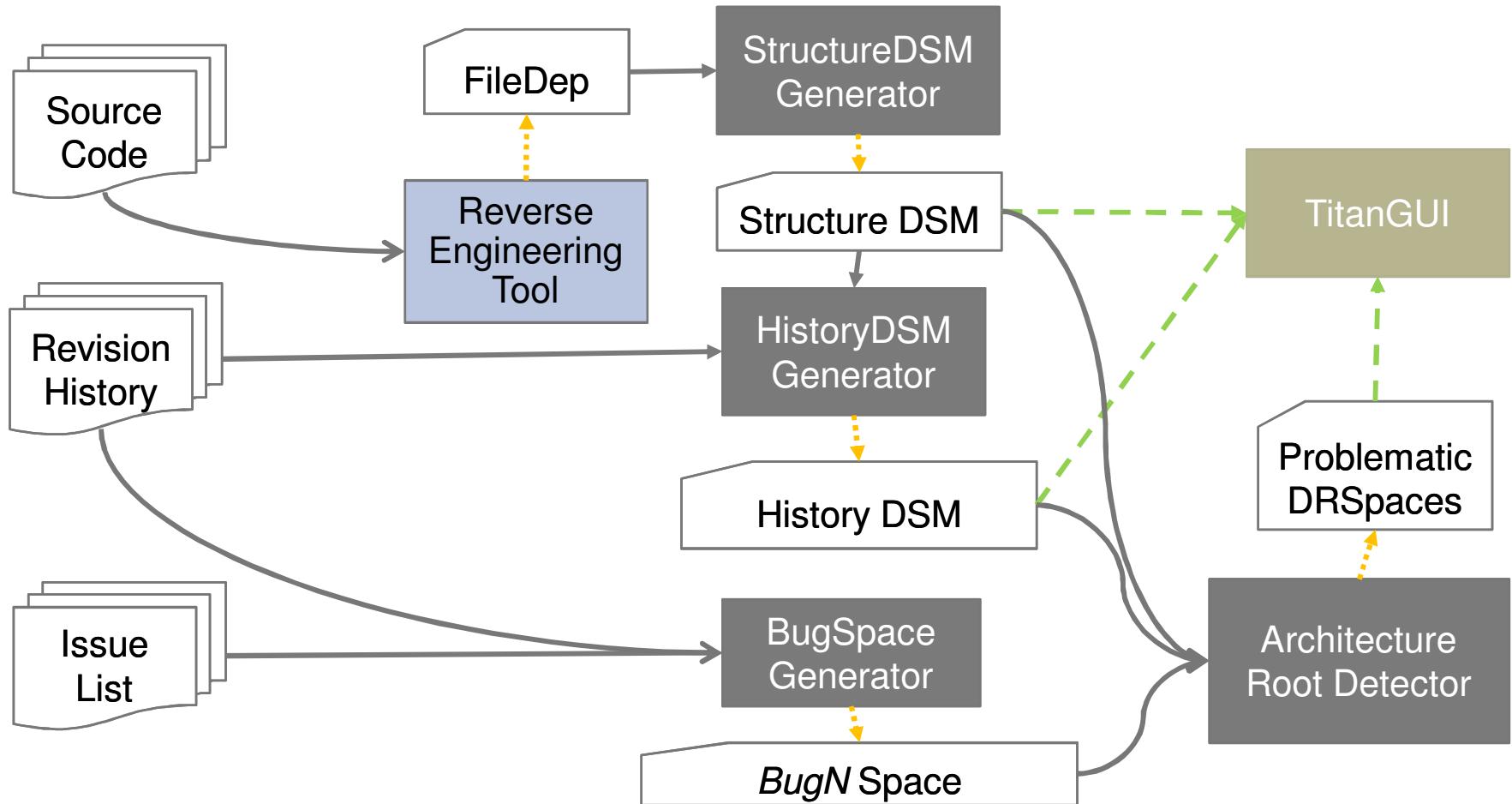
We can view and analyze architectures as a set of (overlapping) DRSpaces.

Each choice of primary relations reveals a different structure.

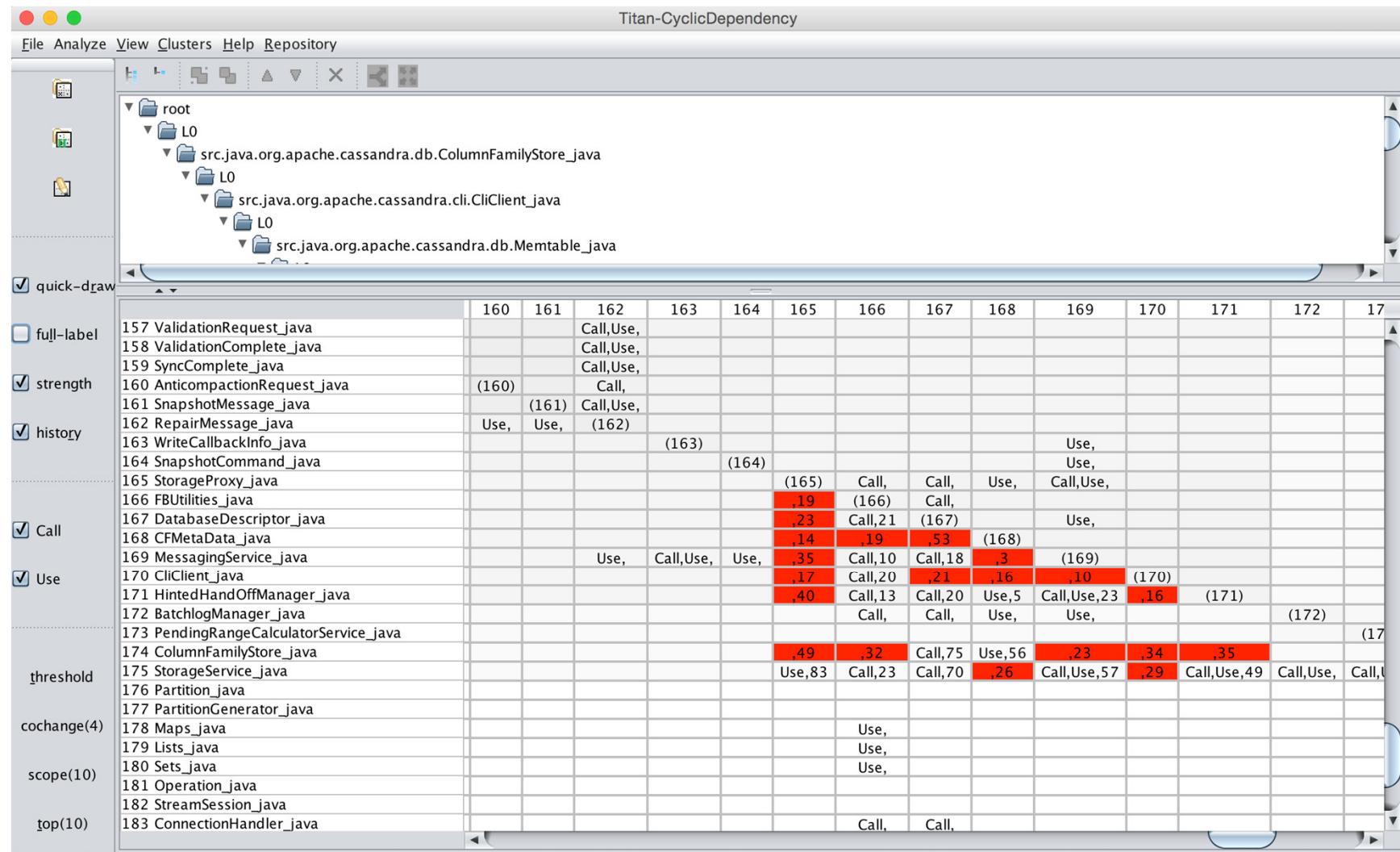


How do we compute DRSpaces?





TITAN GUI

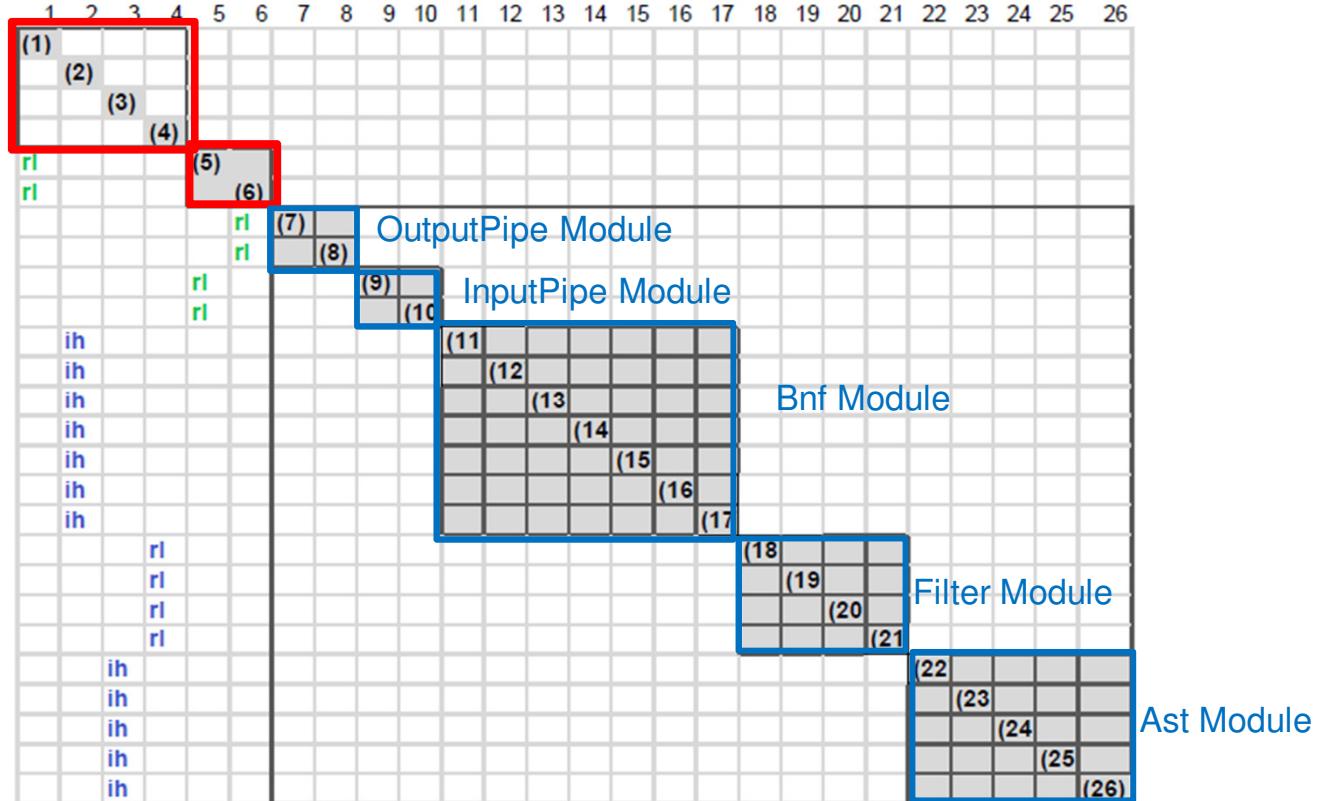


What do different DRSpaces tell us?



REALIZE & INHERIT RELATIONS

```
1 mij.io.Pipe
2 mij.bnf.Node
3 mij.ast.Node
4 mij.Filter
5 mij.io.InputPipe
6 mij.io.OutputPipe
7 mij.io.MemoryOutputPipe
8 mij.io.WriterOutputPipe
9 mij.io.ReaderInputPipe
10 mij.io.MemoryInputPipe
11 mij.bnft.ExponExpr
12 mij.bnft.UnaryExpr
13 mij.bnft.LexExpr
14 mij.bnft.MultExpr
15 mij.bnft.AddExpr
16 mij.bnft.ValueExpr
17 mij.bnft.ParamExpr
18 mij.Interpreter
19 mij.parse.Convert
20 mij.parse.Parser
21 mij.lex.Lexer
22 mij.ast.OpExpr
23 mij.ast.Number
24 mij.ast.FuncExpr
25 mij.ast.Variable
26 mij.ast.UnaryOpExpr
```



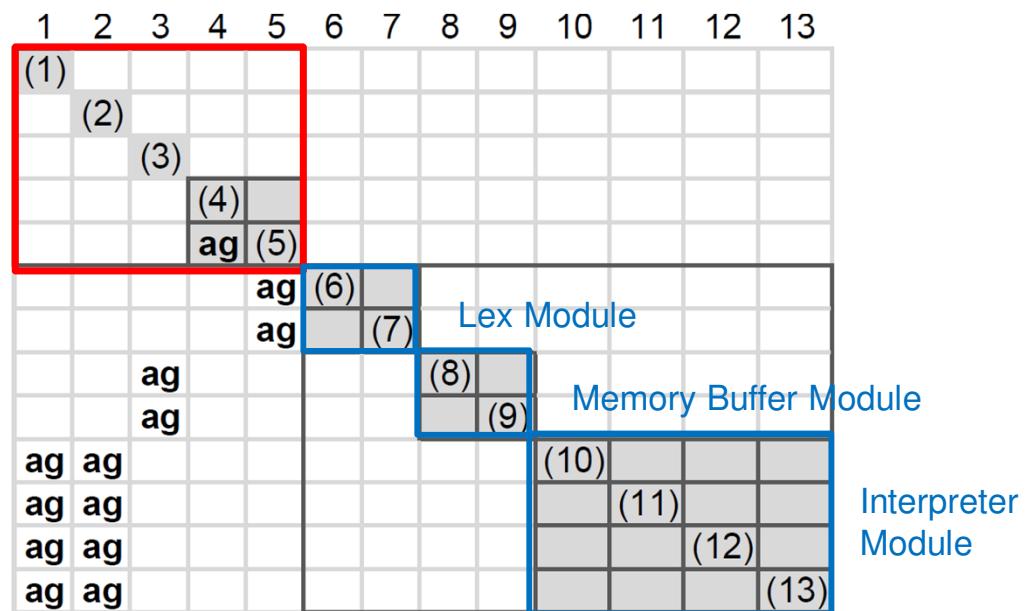
Four leading classes in first layer

InputPipe and OutputPipe in the second layer

These decouple 5 independent modules in the third layer

AGGREGATION RELATION

- 1 mij.io.InputPipe
- 2 mij.io.OutputPipe
- 3 mij.io.MemoryBuffer
- 4 mij.lex.LexType
- 5 mij.lex.Lexeme
- 6 mij.ast.Node
- 7 mij.bnf.LexExpr
- 8 mij.io.MemoryOutputPipe
- 9 mij.io.MemoryInputPipe
- 10 mij.Interpreter
- 11 mij.parse.Convert
- 12 mij.lex.Lexer
- 13 mij.parse.Parser

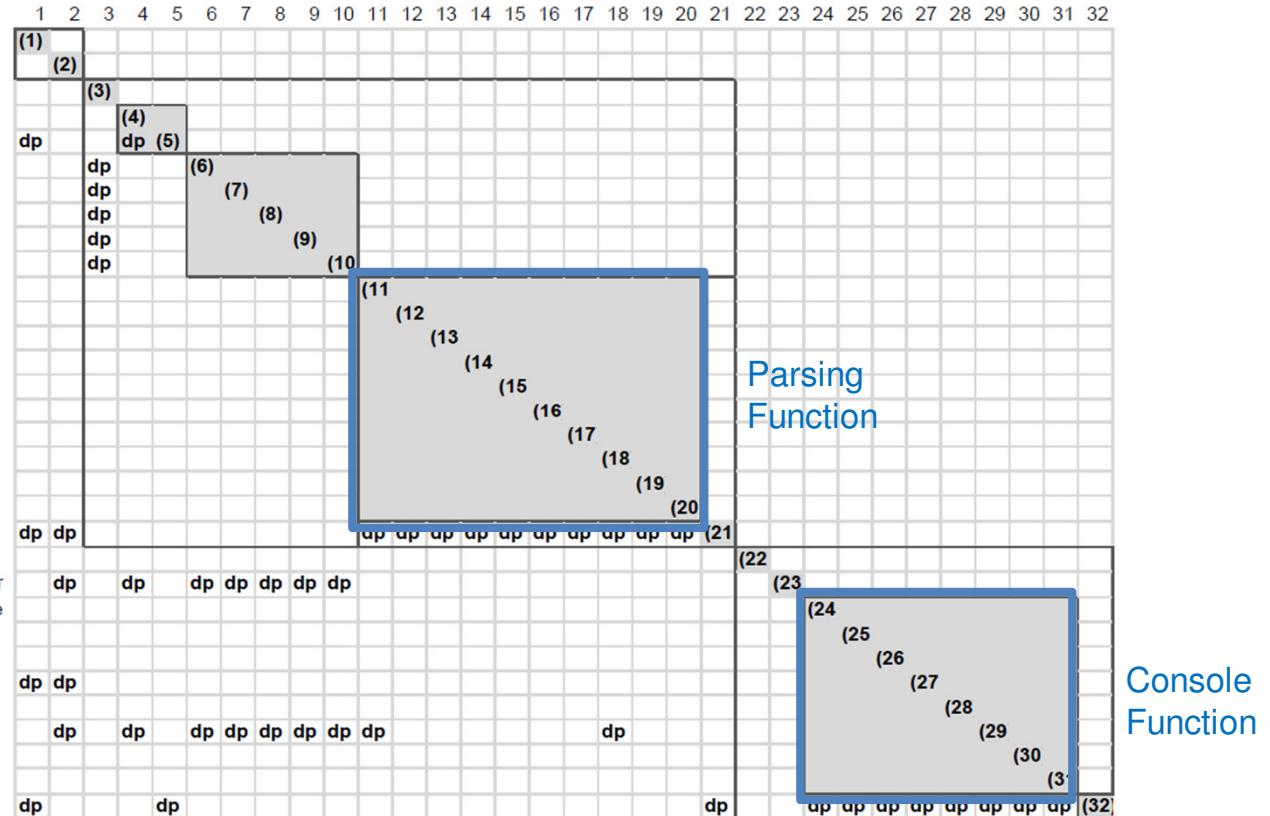


Five leading classes in first layer

Three coherent modules in second layer

DEPENDENCY RELATION

```
1 mij.FilterException
2 mij.lex.Lexeme
3 mij.ast.TreeVisitor
4 mij.ast.Node
5 mij.Interpreter
6 mij.ast.OpExpr
7 mij.ast.Number
8 mij.ast.FuncExpr
9 mij.ast.Variable
10 mij.ast.UnaryOpExpr
11 mij.bnf.LexExpr
12 mij.bnf.ExponExpr
13 mij.bnf.ValueExpr
14 mij.bnf.UnaryExpr
15 mij.bnf.MultExpr
16 mij.bnf.AddExpr
17 mij.bnf.ParamExpr
18 mij.bnf.Node
19 mij.bnf.GrammarType
20 mij.lex.LexType
21 mij.parse.Parser
22 mij.parse.Parser$Entry
23 mij.Interpreter$Calculator
24 mij.io.MemoryOutputPipe
25 mij.io.InputPipe
26 mij.io.MemoryInputPipe
27 mij.lex.Lexer
28 mij.io.MemoryBuffer
29 mij.parse.Convert
30 mij.io.ReaderInputPipe
31 mij.Filter
32 mij.Console
```



How classes work together to accomplish a function

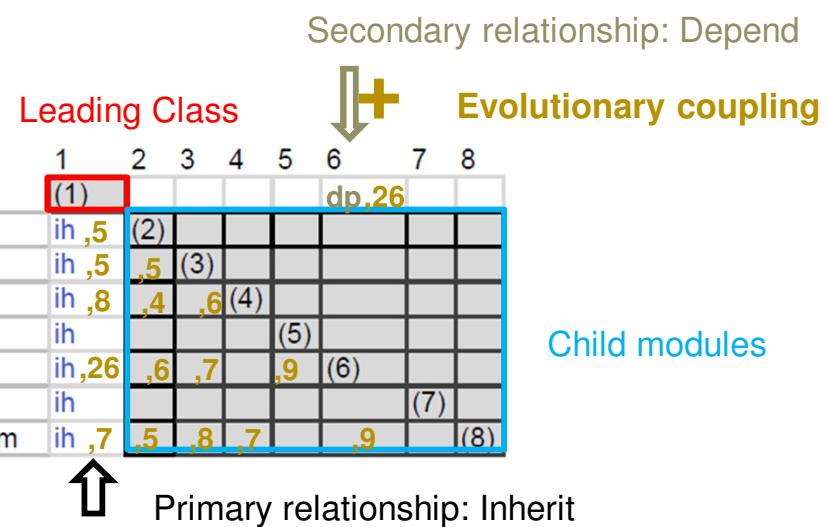
How do we use DRSpaces to reveal architecture hotspots?



REVEALING PROBLEMS

We add in *secondary relationships* to reveal architectural hotspots.

1 org.apache.hadoop.fs.FileSystem								
2 org.apache.hadoop.fs.FilterFileSystem								
3 org.apache.hadoop.fs.RawLocalFileSystem								
4 org.apache.hadoop.fs.s3.S3FileSystem								
5 org.apache.hadoop.fs.kfs.KosmosFileSystem								
6 org.apache.hadoop.dfs.DistributedFileSystem								
7 org.apache.hadoop.dfs.HftpFileSystem								
8 org.apache.hadoop.fs.InMemoryFileSystem\$RawInMemoryFileSystem								



Issue 1: Parent class depends on child class

Issue 2: Unusual evolutionary coupling between parent class and child class

Issue 3: Modularity violation

**Can we find meaningful relationships
between architecture and quality?**



INITIAL RESEARCH QUESTIONS

RQ1: If the leading class of a DRSpace is error-prone, are the files in the DRSpace also error-prone?

RQ2: Are the majority of the error-prone files concentrated in a few DRSpaces?

RQ3: By combining evolution and structure coupling, can we get more insight into architectural problems? Can this help us find not just the locations of errors, but also the reasons for them?

INITIAL EVALUATION DATA SET

Subject	History	#Files	#Releases	#Commits	#Issues
JBoss 3.2.4	Apr 00 – Jun 04	762	11	6458	550
Hadoop 0.15	Feb 06 – Oct 07	692	15	3001	490
Eclipse 3.0.2	May 01 – Mar 05	3,704	10	27,806	3,458

Conventions

- Bug Space: *BugN*
- Design Space bugginess: *dsb*
- Bug Space coverage: *bsc*

RQ1

RQ1: If the leading class of a DRSpace is error-prone, are the files in the DRSpace also error-prone?

RQ1: ANSWER

We selected DRSpaces led by the top 30 most error-prone files as design rules.

We then sub-selected DRSpaces with at least 10 files.

We computed dsb and bsc for each project with respect to Bug2, Bug5 and Bug10.

Consider JBoss: 9 out of 30 DRSpaces have at least 10 files.

62% of the files in each selected DRSpace have more than 2 bug fixes.
The average JBoss DRSpace contains 10% of the project's Bug2 files.

#TopDRS	Avg. in Bug2		Avg. in Bug5		Avg. in Bug10	
	dsb	bsc(#)	dsb	bsc(#)	dsb	bsc(#)
JBoss(9)	62%	10%(206)	50%	13%(129)	18%	30%(23)
Hadoop(11)	67%	11%(187)	53%	14%(106)	47%	18%(68)
Eclipse(11)	31%	7%(291)	15%	9%(98)	9%	13%(36)

RQ1: ANSWER

JBoss:

#DRS	#Bug (Rank)	#Bug2: 206 #(dsb)	bsc	#Bug5: 129 #(dsb)	bsc	#Bug10: 23 #(dsb)	bsc
dr1:18	27(2nd)	10(56%)	5%	7(39%)	5%	4(22%)	17%
dr2:56	21(4th)	32(57%)	16%	26(46%)	20%	10(18%)	43%
dr3:28	18(7th)	22(79%)	11%	18(64%)	14%	5(18%)	22%
dr4:43	14(12th)	27(63%)	13%	20(47%)	16%	5(12%)	22%
dr5:76	11(21st)	43(57%)	21%	35(46%)	27%	12(16%)	52%

- JBoss's Bug2 space has 206 files
- Consider **org.jboss.ejb.Container**:
 - 21 bug fixes; 4th most error-prone file
 - Its DRSpace contains 56 files
- 32 files out of the 56 files have more than 2 bug fixes
 - dsb = 32/56 = 57%
 - bsc = 32/206 = 16%

RQ2

RQ2: Are the majority of the error-prone files concentrated in a few DRSpaces?

RQ2: ANSWER

1) What portion of the bug space can the top 5 and top 10 largest DRSpaces cover?

Proj	Bug2		Bug5		Bug10	
	Top 5	Top 10	Top 5	Top 10	Top 5	Top 10
JBoss	57%	64%	52%	57%	78%	78%
Hadoop	59%	67%	68%	75%	76%	85%
Eclipse	71%	78%	83%	89%	89%	92%

✓ The top 5 and top 10 largest DRSpaces can cover the majority of the files in each bug space.

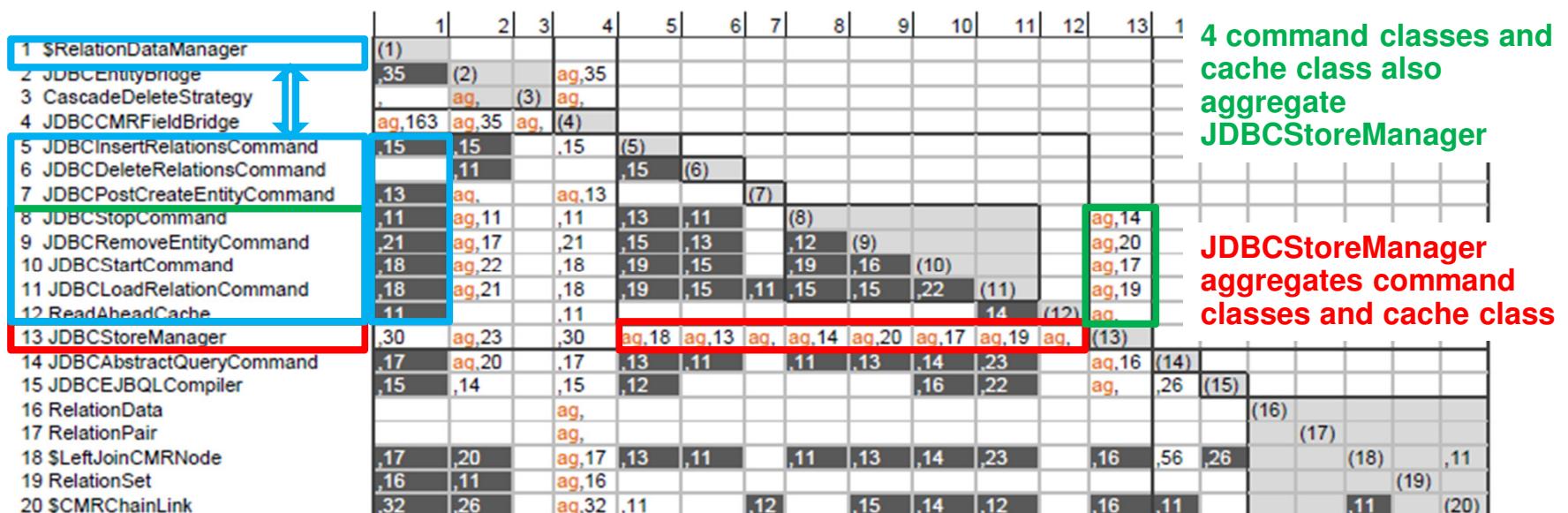
RQ3

RQ3: By combining evolution and structure coupling, can we get more insight into architectural problems? Can this help us find not just the locations of errors, but also the reasons for them?

RQ3: ANSWER

Aggregation cycles

Shared Secrets



Jboss JBCCMRFIELDBridge DRSpace

RQ3: ANSWER

Problematic inheritance Unusual evolutionary coupling

1 org.apache.hadoop.fs.FileSystem

1	2	3	4	5	6	7	8
(1)						dp	26
ih,5	(2)						
ih,5	,5	(3)					
ih,8	,4	,6	(4)				
ih,				(5)			
ih,26	,6	,7	,9		(6)		
ih,						(7)	
ih,7	,5	,8	,7		,9		(8)

2 org.apache.hadoop.fs.FilterFileSystem

3 org.apache.hadoop.fs.RawLocalFileSystem

4 org.apache.hadoop.fs.s3.S3FileSystem

5 org.apache.hadoop.fs.kfs.KosmosFileSystem

6 org.apache.hadoop.dfs.DistributedFileSystem

7 org.apache.hadoop.dfs.HftpFileSystem

8 org.apache.hadoop.fs.InMemoryFileSystem\$RawInMemoryFileSystem

Hadoop FileSystem Inherit DRSpace

ANSWERS TO INITIAL RQS

RQ1: A significant portion of the DRSpaces led by an error prone class are also error-prone.



RQ2: The 5 largest DRSpaces always captured more than half of the files in the bug space.



RQ3: Error-prone DRSpaces also have structural problems and modularity violations.



FOLLOW-ON RQ

RQ4: If a file is involved in greater numbers of architecture issues, it is more error-prone/change-prone than average files?

SECOND DATA SET

Subject	#Month	#Snap	#Comm	#Issue	#File
Avro 1.7.6	47	22	1480	630	145-298
Camel 2.11.1	53	46	17706	2326	528-1203
Cassandra 1.0.7	24	46	6738	3645	419-786
CXF 2.7.10	70	92	27247	3400	1426-3073
Hbase 0.94.16	70	21	14858	5032	347-2142
Ivy 2.3.0	52	11	3799	839	418-607
OpenJPA 2.2.2	68	17	6736	1574	1216-1761
PDFBox 1.8.4	46	13	1798	1279	458-589
Wicket 1.5.5	57	55	18004	3359	1099-1549
Commercial	9	13	6000	800	137-599

RQ4 ANSWER

We counted the architecture issues in the 10 projects and compared these to:

- Bug frequency
- Bug churn
- Change frequency
- Change churn

RQ4 ANSWER

Avro-1.7.6					Camel-2.11.1					Cassandra-1.0.7				
#AI	BF_avg	BC_avg	CF_avg	CC_avg	#AI	BF_avg	BC_avg	CF_avg	CC_avg	#AI	BF_avg	BC_avg	CF_avg	CC_avg
0	0.1	3.7	0.5	29.0	0	0.5	7.9	2.2	58.2	0	0.4	7.1	1.0	32.6
1	0.4	3.9	0.9	26.2	1	1.2	18.5	5.6	131.5	1	1.1	17.4	4.8	106.4
2	1.6	12.6	5.2	376.7	2	3.7	56.6	14.4	304.7	2	5.3	84.5	21.2	559.1
3	7.9	124.5	21.6	628.5	3	8.4	141.5	33.9	681.3	3	12.8	245.8	45.7	1202.0
4	16.5	255.0	33.5	1220.0	4	13.9	204.7	50.9	1043.5	4	18.8	364.9	65.7	1909.4
PC	0.91	0.89	0.94	0.95	PC	0.96	0.96	0.97	0.97	PC	0.97	0.96	0.98	0.97
CXF-2.7.10					Hadoop-2.2.0					HBase-0.94.16				
#AI	BF_avg	BC_avg	CF_avg	CC_avg	#AI	BF_avg	BC_avg	CF_avg	CC_avg	#AI	BF_avg	BC_avg	CF_avg	CC_avg
0	0.8	21.0	2.8	86.9	0	0.4	12.7	1.0	56.8	0	0.7	10.4	0.9	53.0
1	2.9	62.3	9.4	262.5	1	1.5	24.8	4.2	167.7	1	4.8	236.7	8.3	614.6
2	8.6	164.8	23.1	592.0	2	5.3	173.6	13.8	558.3	2	9.9	418.5	17.2	2083.6
3	20.2	390.9	52.5	1232.4	3	26.0	725.1	58.0	1959.6	3	47.8	1335.1	87.6	3158.7
4	54.1	890.2	142.3	3326.0	4	13.7	237.9	26.8	1252.0	4	76.7	2370.4	135.1	6019.0
PC	0.90	0.92	0.89	0.89	PC	0.76	0.63	0.72	0.83	PC	0.93	0.94	0.93	0.97
Ivy-2.3.0					OpenJPA-2.2.2					Pdfbox-1.8.4				
#AI	BF_avg	BC_avg	CF_avg	CC_avg	#AI	BF_avg	BC_avg	CF_avg	CC_avg	#AI	BF_avg	BC_avg	CF_avg	CC_avg
0	0.2	4.5	1.1	31.8	0	1.8	10.0	1.1	36.8	0	0.5	27.1	1.1	92.0
1	1.1	22.8	3.3	79.6	1	3.2	31.1	3.7	111.5	1	1.4	35.9	2.9	136.5
2	2.9	54.6	8.4	251.9	2	4.6	64.5	7.5	229.8	2	1.5	64.1	3.4	259.9
3	7.0	119.9	20.9	646.2	3	10.8	408.6	22.4	862.5	3	8.1	495.0	13.7	861.3
4	6.4	204.6	18.6	792.3	4	25.1	981.0	52.5	2301.1	4	12.2	669.5	18.4	1254.4
PC	0.94	0.96	0.93	0.97	PC	0.90	0.88	0.90	0.88	PC	0.92	0.92	0.94	0.94
Commercial Project														
#AI	BF_avg	BC_avg	CF_avg	CC_avg										
0	0.1	2.25	2.7	102.5										
1	0.2	4.6	5.9	200.4										
2	0.8	3.24	10.3	372.0										
3	2.8	36.8	19.8	884.7										
4	6.0	21	29.0	549.0										
PC	0.91	0.73	0.98	0.81										

AVRO-1.7.6

#AI	BF_avg	BC_avg	CF_avg	CC_avg
0	0.1	3.7	0.5	29.0
1	0.4	3.9	0.9	26.2
2	1.6	12.6	5.2	376.7
3	7.9	124.5	21.6	628.5
4	16.5	255.0	33.5	1220.0
PC	0.91	0.89	0.94	0.95

FOLLOW-ON RQ

RQ4: If a file is involved in greater numbers of architecture issues, it is more error-prone/change-prone than average files?



ECONOMIC ANALYSIS

Based on the identified DRSpaces and a knowledge of their architecture issues, we can determine refactoring strategies.

And we can make decisions about whether to refactor based on ROI.

Consider the following analysis of a commercial project:

ECONOMIC ANALYSIS

	A	B	C	D	E	F	G	I	J	K	L	M	N
1	DRSpace Leading File	DRSpace Size	Norm Size	Current Defects/Yr	Norm Defects	Current Changes/Yr	Norm Changes/Yr	Tot LOC Changed	Norm LOC Changed	Refactor Cost (PM)	Norm Exp Defects/Yr	Norm Exp Changes/Yr	Norm Exp LOC Changed
2	Pear.java	139	103.1	166	123.1	1068	839.2	49,171	36,471	5.5	34	299	17,522
3	Apple.java	158	103.5	63	41.3	607	451.7	25,603	16,772	7	34	300	17,590
4	Bean.java	65	30.4	72	33.7	429	207.2	17,807	8,328	1.5	10	88	5,167
5													
6	DRSpace Total		237		198.1		1498		61,571		78.2	687.3	40,279
7	Project Total		797		265		2332		135,453		14		
8	Savings										120	811	21,292
9													
10													
11	Base defect rates		0.33										
12	Base change rates		2.9										
13	Base LOC/file		169.95										
14	SLOC/PM		600										
15													

Result: >250% ROI in the first year alone!

TAKE-AWAYS

Architectural problems lead to quality issues.

We can locate these problems!

**We can not fix the quality problems without
fixing the underlying
architecture problems!**

**To do this we view and
analyze architectures
as a set of overlapping
DRSpaces.**



TAKE-AWAYS

You can not find these problems by code-based analysis alone.

This analysis allows us to plan refactoring strategies and make informed, *economics-based* decisions about if and how to refactor.





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FURTHER READING

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R. Mo, Y. Cai, R. Kazman, L. Xiao, “Hotspot Patterns: The Formal Definition and Automatic Detection of Architecture Smells”, *Proceedings of The Working IEEE/IFIP Conference on Software Architecture (WICSA 2015)*, (Montreal, Canada), May 2015.

L. Xiao, Y. Cai, R. Kazman, "Design Rule Spaces: A New Form of Architecture Insight", *Proceedings of the International Conference on Software Engineering (ICSE) 2014*, (Hyderabad, India), June 2014.