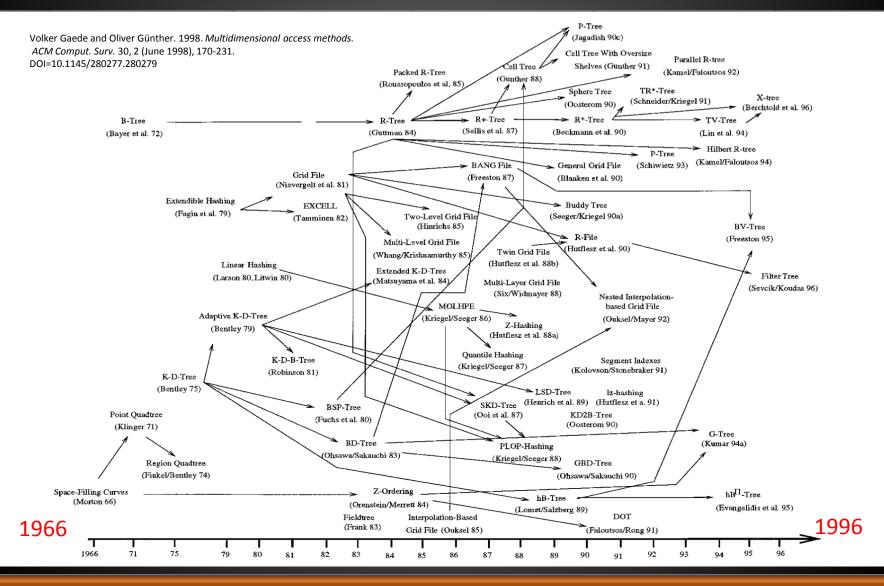
SSDBM, San Diego, USA, 2015

Transparent inclusion, validation, and utilization of main memory domain indexes

Thanh Truong C., Tore Risch Uppsala University Sweden "Many scientific applications involving, e.g., data mining, temporal queries, and spatial analyses, <u>require customized</u> <u>indexing to improve performance</u>"

How many index structures are there ?



How many index structures are used in DBMSs ?

Index structure	ORACLE	My <mark>SQL</mark>	Microsoft SQL Server
Btree	Y	Y	Y
Hash	Υ	Y	Y
R-tree	Y	Y	-
Trie	-	Y	-
Bit-map	Y	-	-

Notes

- Versions
 - Oracle 12c Release 1
 - SQL Server 2014
 - MySQL 5.6
- In MySQL, some storage engines permit some index types, but not all.
- The table does not count "Function based index".
- In SQL Server, hash index is only available for in-memory tables.



In-memory databases

Hao Zhang, Gang Chen, Kian-Lee Tan, Meihui Zhang In-Memory Big Data Management and Processing: A Survey,

Knowledge and Data Engineering, IEEE Transactions on volumn 27, pages 1920 - 1948 DOI: 10.1109/TKDE.2015.2427795

	Systems	Data	Workloads	Indexes
		Model		
	H-Store	relational	OLTP	hashing, B+-
		(row)	0.2.11	tree, binary
				tree
	Hekaton	relational	OLTP	latch-free
		(row)		hashing,
Relational				Bw-tree
Databases	HyPer/	relational	OLTP,	hashing,
	ScyPer		OLAP	balanced
				search tree,
				ART
	SAP	relational,	OLTP,	timeline index,
	HANA	graph, text	OLAP	CSB ⁺ -tree,
	Maria			inverted index
	MemepiC	key-value	object	hashing, skip- list
			operations, analytics	list
	MongoDB	document	object	B-tree
NoSQL	Mongorb	(bson)	operations,	D-uce
Databases		(0501)	analytics	
Databases	RAMCloud	key-value	object op-	hashing
		ney value	erations	intering
	Redis	key-value	object op-	hashing
			erations	Ū
	Bitsy	graph	OLTP	N/A
Graph				
Databases	Trinity	graph	graph op-	N/A
Databases			erations	
	Memcached	key-value	object op-	hashing
Cache	N Cl		erations	1 1
Systems	MemC3	key-value	object op-	hashing
	TxCache	kan usha	erations OLTP	hashing
	M3R	key-value key-value	analytics	hashing N/A
Big Data	Piccolo	key-value	analytics	hashing
Analytics	Spark/	RDD	analytics	N/A
Systems	RDD	KDD	anarytics	N/A
Real-time	Spark	RDD	streaming	N/A
Processing	Streaming			
Systems	Yahoo! S4	Event	streaming	hashing

How many index structures are used in Inmemory databases ?

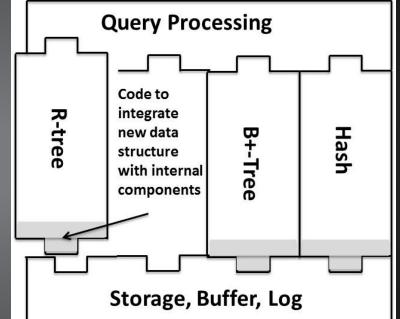
> Hao Zhang, Gang Chen, Kian-Lee Tan, Meihui Zhang In-Memory Big Data Management and Processing: A Survey,

Knowledge and Data Engineering, IEEE Transactions on volumn 27, pages 1920 - 1948 DOI: 10.1109/TKDE.2015.2427795 Why?

Because it is very challenging

Here are some challenges C1,..,C5

- C1. Understanding the DB kernel
- C2. Re-implementing the datastructure
- C3. Integrating with other DB internal components
- C4. Extending query processor
- C5. Validating the index's functionalities



Only database kernel) expert can do it

Solution?

Some extensible indexing frameworks

• GiST

J Hellerstein. M., J.F. Naughton, and A. Pfeffer: *Generalized search trees for database systems*, Proc. VLDB Conf., pp 562–573, 1995.

• Extensible Indexing – Orcale 8i

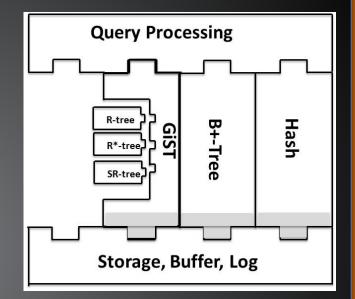
J. Srinivasan, R. Murthy, S. Sundara, N. Agarwal, and S. DeFazio: *Extensible indexing: a framework for integrating domain-specific indexing schemes into oracle8i*. Proc. ICDE Conf., pp 91–100, 2000.

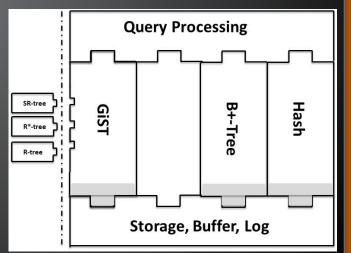
SP-GisT

W. G. Aref and I. F. Ilyas: *An extensible index for spatial databases*, Proc. SSDBM, pp 49–58, 2005.

Reviews

- These frameworks specifies coding conventions and primitives.
- Solved C1 Understanding DB kernel -
- Solved C3 Integrating with other kernel components





The remaining unsolved challenges

- C2 Re-implementing the index implentation It is not OK if the index implementation
 - has ownership.
 - Is available in binary.
 - or being very complex to re-implement, i.e; Judy-tries
 ????
- C4 Extending query processor ????
- C5 Validating the index's functionalities ????

Our motto

Only database (kernel) expert can do it!

"It should not be necessary to be a database kernel expert to introduce a new domain index"

Our solution

• The paper title:

"Transparent inclusion, utilization, and validation of main memory domain indexes"

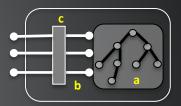
- The paper itself
 - Transparent inclusion to solve C1, C2, C3
 - no index implementation code changed .
 - Transparent utilization to solve C4
 - automatically transforms queries to utilize the new added index.
 - Transparent validation to solve C5
 - Automatically generates and executes queries to test the new added index
- The result :

The generalized extensible indexing framework: Main-memory eXternal Index Manager (Mexima).

Website: <u>http://www.it.uu.se/research/group/udbl/mexima/</u>

How to introduce a new index ?

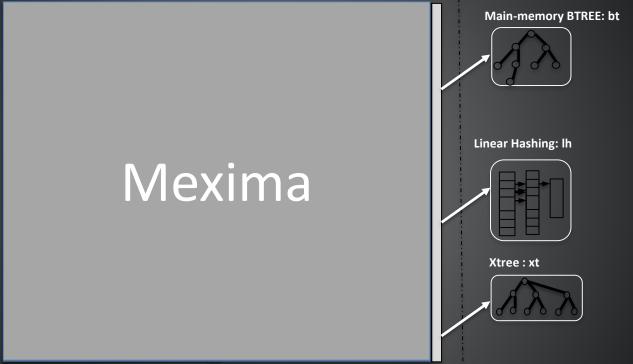
- Grab the *index implementation* (a)
- Study the public *index API* (b)
- Write the *index driver* (c)



(glue code) that interfaces Mexima and the index API

Compiled as dynamic library called as *index* extension

At the end of the day



Windows: dynamic libaries Unix/OSX: shared objects

/* Load main-memory BTREE index*/
load_extension("bt");

At the end of the day (cont.)

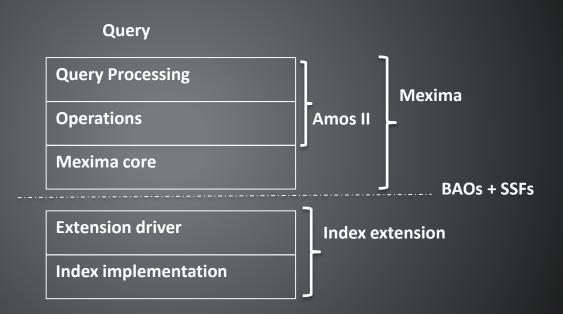
/* Create a table to store salaries of people given social security numbers*/
create function salary(Number ssn)->Number sl as stored;

/*create BTREE on sl*/
create_index("salary", "sl", "BTREE", ");

/*Add data*/
set salary(8301318971) = 2000;
set salary(8501332978) = 3000;

set salary(8001335978) = 4000; /*Query*/ SELECT ssn, sl FROM Number ssn, Number sl WHERE salary(ssn) = sl AND sl >= 3000;

Mexima



Mexima interface = BAOs + SSFs

The index driver code contains
Basic access operators (BAOs) *– create(), drop(), put(), delete(), get(),*

and *map()* that scans the index by applying a specified mapper function on each index entry.

implemented as C functions(***details in the paper)

The index driver code contains

- Special search functions (SSFs)
 - Examples:
 - interval search on B-trees: bt_select_range()
 - and proximity search on X-trees/R-trees: xt_proximity_search()
 - and KNN search on X-trees/R-trees:
 xt_knn_search()
 - Implemented as foreign functions (UDFs)

*** details in the paper)

But it is not enough ...

- How new index is utilized in query?
 - Option 1

End-user can manually call a SSF in query by reformulating the query

• Option 2 End-user can express query naturally, but the query optimizer should be able to utilize the index.

The query processor should transparently transform the query to SSF if possible to utilize the index → SSF translation rules.

SSF translation rules

- An SSF translation rule describes how query fragments are translated to a new format to <u>expose SSFs</u>.
- Examples

#	Index type	priority		Relation	SSF
· ·				operators	
			function		
1	B-tree	1	Nil	>=, <=	btree_select_range
2	X-tree	1	distance	<=	xt_proximity_search
3	X-tree	2	Knn	nil	xt_knn_search

Example - Table

- Table images(id, hist)
 - Id, image's identifier
 - hist, histogram as image's feature vector

Btree index

Xtree index

Example - Query 1

"Query 1 finds images q whose identifiers are between 30 and 100"



With Mexima, it is done by the following SSF translation rule

#	Index type	priority	Index sensitive function	Relation operators	SSF	pf
1	<i>B-tree</i>	1	Nil	>=, <=	btree_select_range	F
	X-tree	1	distance	<=	xt_proximity_search	
	X-tree	2	Knn	nil	xt_knn_search	

	IndexIndextypesensitivefunctionB-treeNil	Relation operators >=, <=	SSF btree_select_r	uery 1 range een 30 and 10	
Input que Q1(q):- images(q, _) q >= 30 q <= 100			Intermed TQ1(q):- (q,_) in btree_selec	iate query t_range(#'images', 0, 30,100)	
Form (i): P(iv,) AND	(iv r1 expression) (iv r2 expression)	AND AND			
	 (iv rn expression)	e f			
Here,					
	riable bound to an table $P()$. We say riable.				
\Box r_i are com	parison operators i	n the set			
	<i>relop</i> , where <i>relop</i>				

Example - Query 2

"For a given image x find the images q whose feature vectors are closer than epsilon (eps = 0.11)."

Input query

Q2(x, q):-

images(x, hist_x) AND images(q, hist_q) AND distance (hist_x, hist_q) <= 0.11</pre>

MAP

Intermediate query

TQ2(x, q):-
image(x, hist_x)AND
(q,hist_q) in
xtree_proximity_search(#'images',1, hist_x, 0.11)AND
distance (hist_x, hist_q) <= 0.11</th>

SSF X-tree proximity search

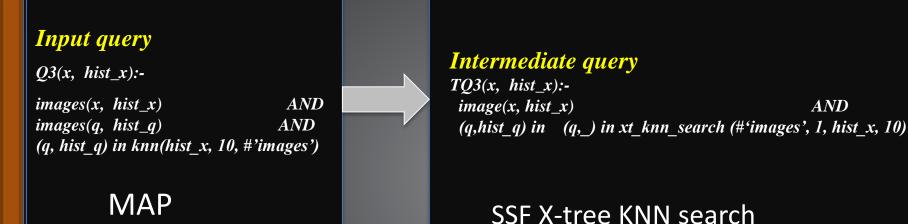
With Mexima, it is done by the following SSF translation rule

#	Index type	priority	Index sensitive function	Relation operators	SSF	pf
	B-tree	1	Nil	>=, <=	btree_select_range	F
2	X-tree	1	distance	<=	xt_proximity_search	T
	X-tree	2	Knn	nil	xt_knn_search	

Index type	Index sensitive function	Relation operators	e - Qi	Jery 2	
" For a given im	<i>distance</i> lage <i>x</i> find i	<= the images (xt_proximit	y_search ture vectors are closer tha	
Input query Q2(x, q):- images(x, hist_x) images(q, hist_q) distance (hist_x, hist_	$AND \\ AND \\ q) <= 0.11$				AND AND
<u>Form (ii):</u> P(iv,) AND isf(iv) r - erni	ression AND	SSF X-		
isf(.,iv,) r ₂ expr .,iv,) r ₂ expr	ession AND	wing SS		
Here, <i>iv</i> is an i parameter position <i>isf()</i> .		-	a torra		
2 X-tree	1 I	listance <	<=, <=		

Example - Query 3

"Find the k = 10 closest images compared to a given image x"



With Mexima, it is done by the following SSF translation rule

#	Index type	priority	Index sensitive function	Relation operators	SSF	pf
	B-tree	1	Nil	>=, <=	btree_select_range	F
	X-tree	1	distance	<=	xt_proximity_search	Т
3	X-tree	2	Knn	nil	xt_knn_search	F

	Index typ	e Index sensitive function <u>Knn</u>	- I	s SSF xt_knn_	to a given image x ²	
3(x, nages	t query hist_x):- s(x, hist_x)		AND	TQ3(x, hist image(x, h		AND ages'. 1. hist x.
, hist For	s(q, hist_q) t_q) in knn(hist_ m (iii): ,iv,) AND	_x, 10, #'ima	ges')		X-tree KNN search	.geo (1, 1131_1.)
, hist For P(.	t_q) in knn(hist m (iii): ,iv,) AND	_x, 10, #'ima _s (,iv,) in isf	ges') f(,P,)	SSF		.geo y 2,,
, hist For P(.	t_q) in knn(hist m (iii): ,iv,) AND	_x, 10, #'ima _s (,iv,) in isf	ges') f(,P,)	SSF	X-tree KNN search	,geo y 2, mor_,y
, hist For P(.	t_q) in knn(hist m (iii): ,iv,) AND	_x, 10, #'ima _s (,iv,) in isf	ges') f(,P,) done by the Index sensitive	SSF e following Relation	X-tree KNN search SSF translation rule	
, hist For P(.	t_q) in knn(hist m (iii): ,iv,) AND	_x, 10, #'ima _s (,iv,) in isf	ges') f(,P,) done by the Index sensitive	SSF e following Relation	X-tree KNN search SSF translation rule	

Reviews of query fragment forms

Form (i):		
P(iv,) AND	(iv r ₁ expression)	AND
	(iv r ₂ expression)	AND
	\dots (iv r_n expression)	
Form (ii):		
P(iv,) AND	isf(,iv,) r ₁ expression	AND
	isf(,iv,) r ₂ expression	AND
	•••	
	isf(,iv,) r _n expression	
Form (iii): P(,iv,) AND (,iv,) in isf(,P,,)
Form (iv): P(.iv,) AND F(iv) relop expression	

Form (v): P(...,iv,...) AND F(isf(...,iv, ...)) relop expression

Mexima

Transparently transformation to utilize indexing

** T. Truong, T. Risch: Scalable Numerical Queries by Algebraic Inequality Transformations, Proc. Database Systems for Advanced Applications (DASFAA), pp 95-109, 2014 isf(...) relop expression

isf(...) LIKE expression

Oracle

Advisor tools to suggest on reformulating the query to utilize indexing

- D. Benoit, D. Das, K. Dias, K. Yagoub, M. Zait, and M. Ziauddin: Automatic SQL tuning in Oracle 10g, Proc. VLDB Conf, pp 1098-1109, 2004.
- Oracle Inc: *Query Optimization in Oracle Database 10g Release 2.*

http://www.oracle.com/technetwork/database/bidatawarehousing/twp-general-query-optimization-10gr-130948.pdf , 2005

Our solution

• The paper title:

"Transparent inclusion, utilization, and validation of main memory domain indexes"

- The paper itself
 - Transparent inclusion
 - o no index implementation code changed .
 - ✤ <u>Transparent utilization</u>
 - o automatically transforms queries to utilize the new added index.

✤ Transparent validation

 Automatically generates and executes queries to test the new added index

What to test ?

BAOs: correctness of BAOs

• SSFs

Correctness of SSFs
Correctness of SSF translation rules

BAO tester

 Automatically tests correctness of put(), get(), delete(), map(), and drop().

Index key generator as queries

#	Idxtype	Index key type	Index KeyGenerators
1	B-tree	Number	select uniform_int(1000,0,10000)
2	X-tree	Vector-Number	select uniform_vec_real(1000,5,0,1)
3	X-tree	Vector-Number	select
			CSV_file_rows("colorhistogram.csv")

BAO tester (cont.)

- Populate generated data into
 - Table I_Table(k, v), having index to test at column k
 - Table R_Table(k,v), having Hash index at k

Execute BAO tester algorithms (***)

• Validate I_Table against R_Table

*** details in the paper

SSF tester - Ideas

- Create sample tables with and without the index
- Auto-generate validation queries
- Recall, SSF translation rules transform these queries

Same value returned if there is no index, or no matching SSF translation rules

SSF tester – Ideas (cont.)

• SSF parameter generators

#	Index type	SSF name	SSF parameter generator	SSF Parameter types
1	B-tree	btree_select_range	select l, u from Number l, Number u where l in uniform_int(100, 0,10000) and u in uniform_int(100,0,10000)	(Number, Number)
2	B-tree	btree_select_open	select u from Number u where u in uniform_int(100, 0,10000)	(Number)
3	X-tree	xtree-proximity-search.	select x, d from Vector of Number x, Number d where x in uniform_vec_real(100,5,0,1) and d in uniform_real(100,0, 1.4)	(Vector of Number, Number)
4	X-tree	xtree_knn-search	select x, k from Vector of Number x, Number k where x in uniform_vec_real(100,5,0,1) and k in uniform_int(0,5)	(Vector of Number, Number)

It is getting more complicated!

SSF tester – Ideas (cont.)

Join three index property tables

– SSF translation rule table

#	Index type	priority	Index sensitive	Relation	SSF	pf
			function	operators		
1	B-tree	1	Nil	>=, <=	btree_select_range	F
2	X-tree	1	distance	<=	xt_proximity_search	Τ
3	X-tree	2	Knn	nil	xt_knn_search	F

Index key generator table

#	Idxtype	Index key type	Index KeyGenerators		
1	B-tree	Number	select uniform_int(1000,0,10000)		
2	X-tree	Vector-Number	select uniform_vec_real(1000,5,0,1)		
3	X-tree	Vector-Number	select CSV_file_rows("colorhistogram.csv")		

– SSF parameter generator table

#	Index	SSF name	SSF parameter generator	SSF	Parameter
	type			types	
1	B-tree	btree_select_range	select l, u from Number l, Number u where l in uniform_int(100, 0,10000) and u in uniform_int(100,0,10000)	(Number	, Number)

SSF tester – Validation queries

• Validation query matching Form (I)

Formula	Descriptions				
select iv, v from IT iv, Number v, T1 p1, T2 p2,, Tm pm where I_Table(iv, v) and (p1, p2,,pm) in (SPG) and (iv r1 p1) and (iv r2 p2) and (iv rm pm); Example select iv, v from Number iv, Number v, Number p1, Nur	 Here, SSF parameters types in the SSF parameter generator table are T₁, T_m (Table 3) <i>IT</i> is the index key type in the index key generator table (Table 2) <i>SPG</i> is the SSF parameter generator (Table 3) for parameters p₁, p_m, and that r_i are the relops in Form (i). 				
where I_Table(iv, v) and (p1, p2) in (select I, u from Number I, Number u where I in uniform_int(100, 0,10000) and u in uniform_int(100,0,10000)) and iv >= p1 and iv <= p2;					

SSF tester – Validation queries

• Validation queries matching Form (II)

Formula	Descriptions			
select iv, v	Here,			
from IT iv, Number v,	• <i>j</i> is the arity of ISF()			
$T_1 p_1, T_2 p_2,, T_m p_m,$				
T_i res				
where I_table(iv, v) and				
$(p_1, p_2,, p_m)$ in (SPG) and				
res = ISF (iv, p_1, \dots, p_{i-1}) and				
$(\operatorname{res} r_1 p_j)$ and				
$(\operatorname{res} r_{\mathrm{m}} p_{\mathrm{m}});$				
Example				
select iv, v from Vector of Number iv, Number v, Vector of Number p1, Number p2, Number res where I_Table(iv, v) and (p1, p2) in (select x, d from Number x, Number d where x in uniform_vec_real(100,5,0,1) and d in uniform_real(100,0, 1.4)) and res = distance(iv, p1) and res<= p2;				

SSF tester – Validation queries

• Validation queries matching Form (III)

Formula
select iv, v
from IT iv, Number v,
$T_1 p_1, T_2 p_2,, T_m p_m,$
where I_table(iv,v) and
$(p_1, p_2,, p_m)$ in (SPG) and
(iv,v) in ISF $(p_1,,p_m, I_Table)$
Example
select iv, v
from Vector of Number iv, Number v, Vector of Number p1,
Number p2
where I_Table(iv, v) and
(p1, p2) in (select x, k from Number x, Number k
where x in uniform_vec_real(100,5,0,1) and k in
uniform_int(0,5)) and
(iv,v) in knn(p1, p2, #'images');

Experiments

Experiment - Purposes

• Code size

- Compare coding size to introduce some indexes between Mexima vs other extensible indexing frameworks
- To show Mexima requires no code change, driver code (glue code) is small
- BAO overhead
 - Time to run a stand-alone index implementation
 - Time to run it when plugging into Mexima
 - To investigate the overhead = Penalty of using Mexima
- Impact of SSF translation rules
 - Time to run queries with/without SSF translation rules
 - To show the importance of query rewrite to utilize indexes

Experiment - Settings

 All performance experiments were repeated 10 times, from which the average figures were calculated after removing outlier results if any.

 The experiments were run under Windows 7 on an Intel (R) Core(TM) i5 760 @2.80GHz
 2.93 GHz CPU with 8GB RAM, using the Visual Studio 10 32 bits C compiler.

Experiment – Code size

 Count number of code C/C++ lines of glue code vs other extensible indexing systems

	GiST	SP-GiST	Mexima	Factor
B-tree	5031		116	43
KD-tree		572	118	5
R-tree	1133		120	9.5
Trie		580	120	5

- PostgreSQL version 9.3.5, <u>http://www.postgresql.org/ftp/source/v9.3.5/</u>
- SP-GiST version 0.0.1, <u>https://www.cs.purdue.edu/spgist/</u>
- o Mexima http://www.it.uu.se/research/group/udbl/mexima

• Mexima requires

- no code change to the index implentation
- litle coding effort for the driver (interface)

Experiment – BAO overhead

• The total time

tot = *op* + *mc* + *ed* + *st*

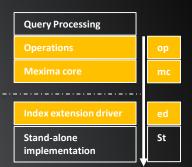
- The Mexima overhead,
 o= op + mc + ed
- Breaking down the overhead
 - %op ?
 - %mc?



Query Processing	
Operations	ор
Mexima core	mc
Index extension driver	ed
Stand-alone implementation	St

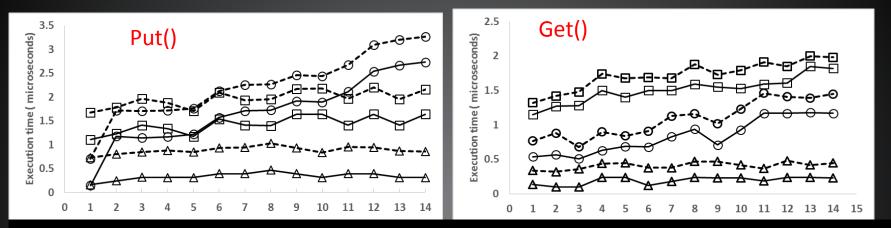
Experiment – BAO overhead (cont.)

BAO	Index	0	%ор	%mc	%ed
Put	LH	0.56	51.7%	36.2%	12.1%
	B-tree	0.53	52.3%	35.8%	11.9%
	Judy-trie	0.54	52%	35.3%	11.7%
Get	LH	0.26	37.2%	47.1%	15.7%
	B-tree	0.23	36.6%	47.6%	15.7%
	Judy-trie	0.22	36%	48%	16%
Мар	LH	0.07	32.1%	50.9%	17%
	B-tree	0.07	34.4%	49.2%	16.4
	Judy-trie	0.07	33.7%	49.7%	16.6%
Delete	LH	0.42	45%	41.3%	13.7%
	B-tree	0.42	43.3%	42.5%	14.2%
	Judy-trie	0.41	43.4%	42.5%	14.1%

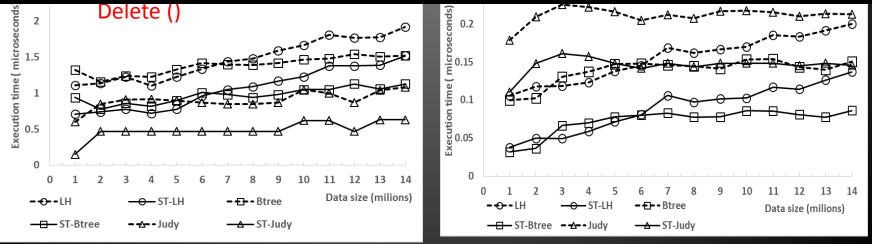


- Data size = 5 milion key/value pairs
- 1000 random inserts, lookups, deletes.
- The average overhead in microseconds per call
- Overhead < 0.6 microsecond</p>

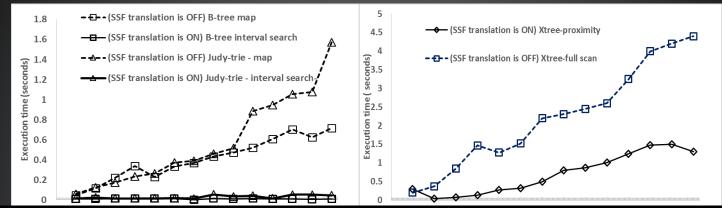
Experiment – Overhead w.r.t data size



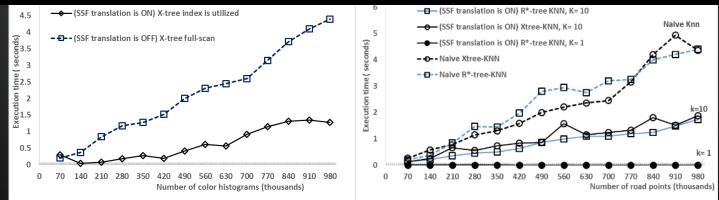
The stand-alone index implementations are always faster The overhead is independent of the database size



Experiment – Impact of SSF translation rules



With SSF translation rules, queries run faster. It made indexes utilized



Xt-trees on Proximity search

X-trees and R*-trees on KNN search

Experiment – Side notes

- Bugs found in the following used index implementations using Mexima tester
 - X-trees [1]
 - R*-trees [2]
 - B-trees [1]
- Comparisons
 - Judy-trie [ref] outperformed B-trees in get(), insert(),delete(), but not map()
 - For 2D 4D, X-trees is as good as R*-trees
 - For higher dimension (9D), X-trees is applicable and scale

[1] http://www.it.uu.se/research/group/udbl/mexima[2] http://www.ics.uci.edu/~salsubai/rstartree.html

Conclusions & Future work

Conclusions

- The Mexima framework allows plugging-in of mainmemory domain index implementations with ease
 - without code changes
 - a simple Mexima driver for BAOs and SSFs
 - declare index properties as queries
 - transparently, Mexima makes new indexes utilized
 - automatically generating and executing validation queries, Mexima validates correctness of BAOs and SSFs
- Tool for testing and comparing indexes
- Future work
 - More indexes will be plugged-in
 - It might put additional requirement to Mexima

Thank you!