

Turbocharging Data Visualization & Analyses
With Oracle In-Memory 12.2
Dr.-Ing. Holger Friedrich



Turbocharging Data Visualization and Analyses with Oracle In-Memory 12.2

Dr.-Ing. Holger Friedrich, CTO
sumIT AG

- Introduction
- Analytic Workloads & Requirements
- Oracle In-Memory (12.2)
- Conclusions

- Consulting and implementation services in Switzerland
- Experts for
 - Data Warehousing,
 - Business Intelligence,
 - Advanced Analytics
- Focussed on Oracle technology
- Oracle specialisations for BI Foundation, Database, Data Warehousing, and Data Integration
- Our motto: Get Value From Data
- Visit our web site: www.sumit.ch (in German)



Specialized
Oracle Business Intelligence
Foundation



Specialized
Data Warehousing



- Computer Science diploma of Karlsruhe Institute of Technology (KIT)
- Ph.D. in Robotics and Machine Learning
- More than 17 years experience with Oracle technology
- Expert for
 - Data Integration
 - Data Warehousing / Big Data,
 - Advanced Analytics and
 - Business Intelligence
- Technical Director of sumIT AG



 Oracle ACE for DWH/BI



Cloud





Big Data



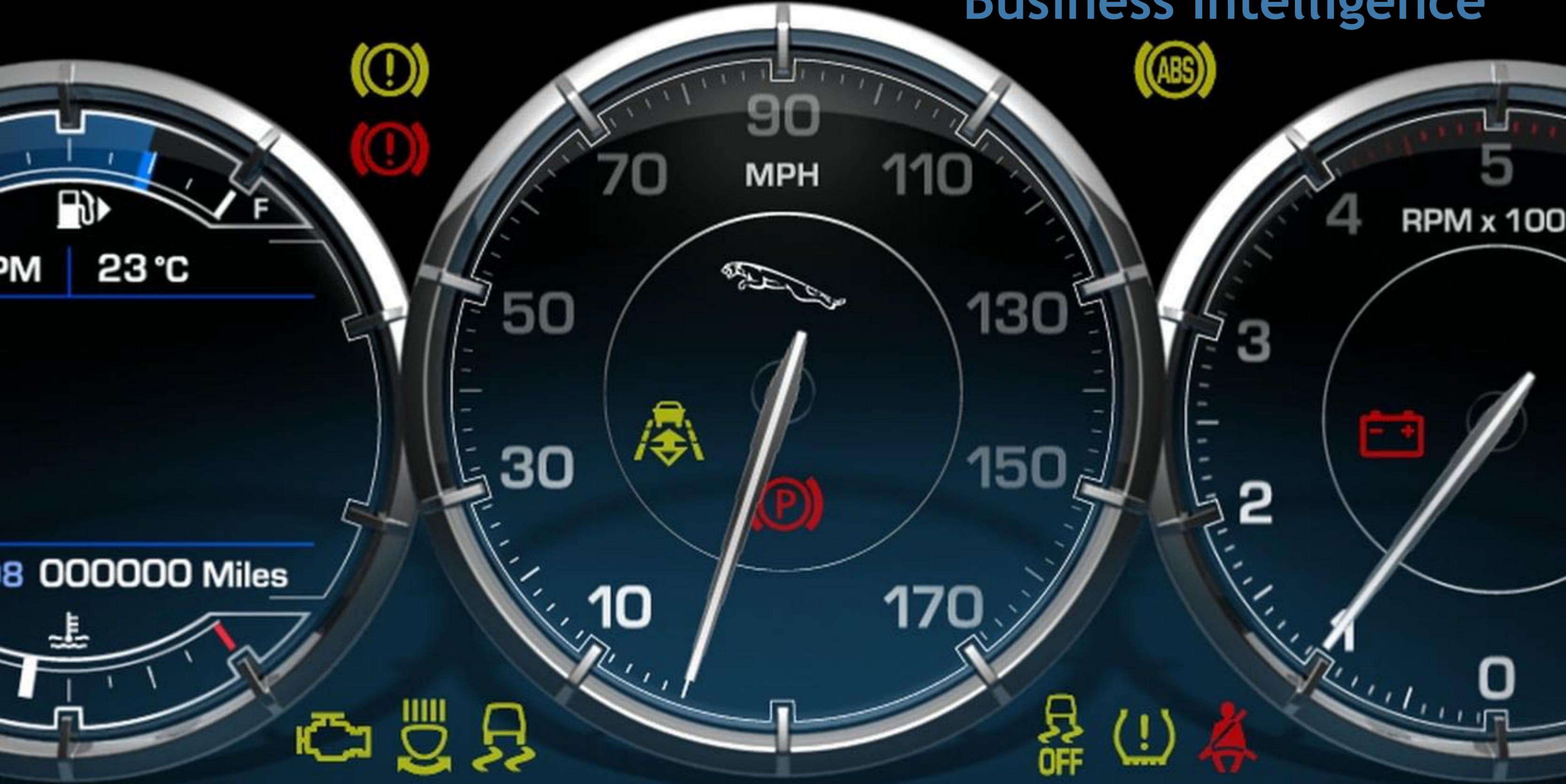
Batch



Streaming



Business Intelligence



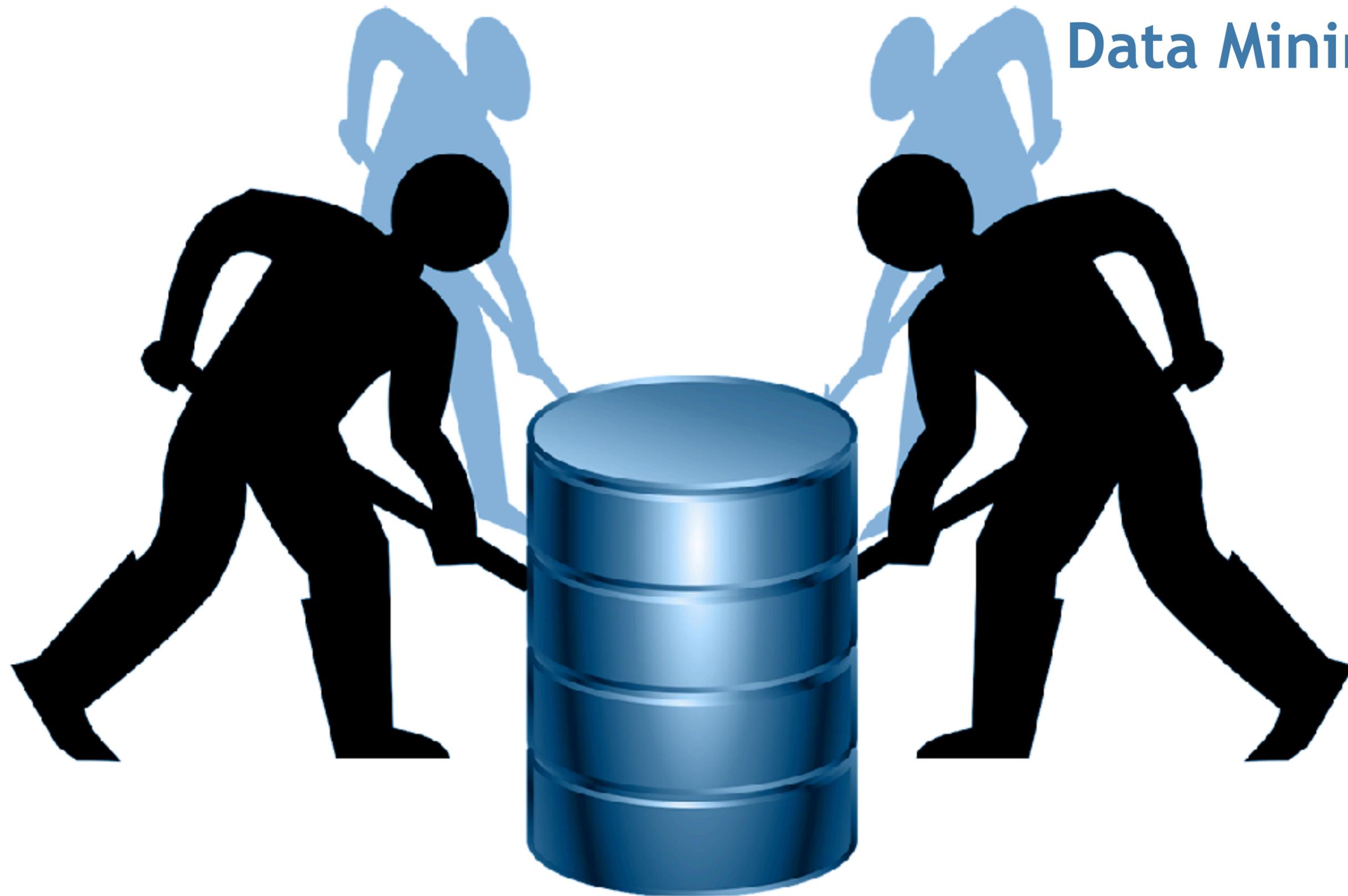
Type (disease/gene)

Nodes

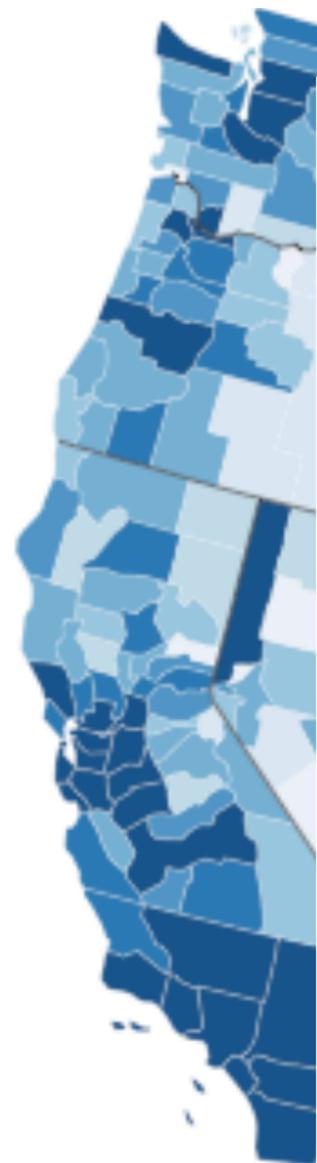
Betweenness

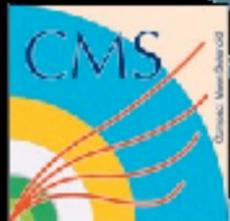


Data Mining



fea
S



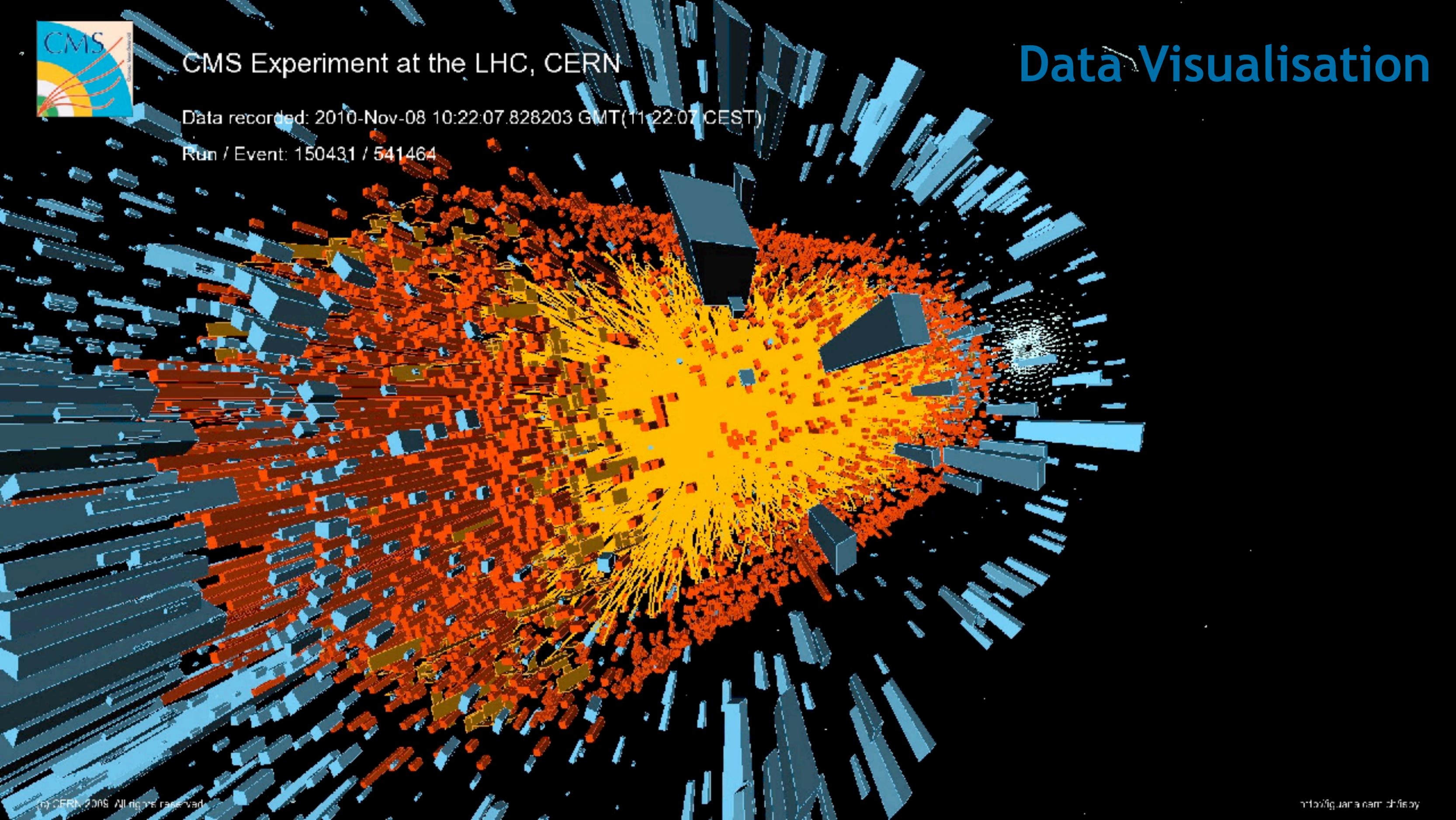


CMS Experiment at the LHC, CERN

Data recorded: 2010-Nov-08 10:22:07.828203 GMT(11-22:07 CEST)

Run / Event: 150431 / 541464

Data Visualisation



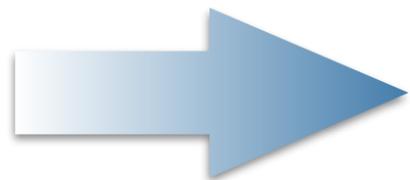
Data Curation



Single Source of Truth



- Scans of large amounts of data
- Apply functions to single columns
- Compute complex expressions on multiple attributes,
- Aggregate data
- Join data sets through known attribute pairs
- Flexibly join data sets through arbitrary attribute pairs
- Data sets do range from small to very large
- Query sensitive data sets
- Query new data sets (schema on read, JSON, etc.)



- All of the above with high speed

- Best for queries that
 - scan large quantities of data
 - compute aggregates on the results
- Efficient storage use, due to high compression benefits on most columns



1. In-memory storage indexes
2. Filtering on binary compressed data
3. Columnar storage of selected columns
4. Transparent querying across storage hierarchy
5. Real-time background synching of columnar store
6. Parallel query execution on the columnar store
7. SIMD vector processing
8. In-memory fault tolerance on RAC
9. In-memory aggregation



```

SELECT
  /*+ no_parallel(t) no_parallel_index(t) dbms_stats cursor_sharing_exact
      use_weak_name_resl dynamic_sampling(0) no_monitoring
      xmlindex_sel_idx_tbl no_substrb_pad */
  COUNT(*),
  SUM(sys_op_opnsize("PROD_ID")),
  SUM(sys_op_opnsize("CUST_ID")),
  COUNT(DISTINCT "CHANNEL_ID"),
  SUM(sys_op_opnsize("CHANNEL_ID")),
  substrb(dump(MIN("CHANNEL_ID"),16,0,64),1,240),
  substrb(dump(MAX("CHANNEL_ID"),16,0,64),1,240),
  COUNT(DISTINCT "PROMO_ID"),
  SUM(sys_op_opnsize("PROMO_ID")),
  substrb(dump(MIN("PROMO_ID"),16,0,64),1,240),
  substrb(dump(MAX("PROMO_ID"),16,0,64),1,240),
  COUNT(DISTINCT "QUANTITY_SOLD"),
  SUM(sys_op_opnsize("QUANTITY_SOLD")),
  substrb(dump(MIN("QUANTITY_SOLD"),16,0,64),1,240),
  substrb(dump(MAX("QUANTITY_SOLD"),16,0,64),1,240),
  SUM(sys_op_opnsize("AMOUNT_SOLD"))
FROM "SH"."MY_SALES" t;

```

4. aggregations

1. scanning columns

3. post-processing scanned columns

2. single table scan

- In-Memory**

1. In-Memory chosen automatically

Id	Operation	Name	Rows	Bytes	Cost (%CPU)	Time
0	SELECT STATEMENT		1	24	1070 (27)	00:00:01
1	SORT GROUP BY		1	24		
2	TABLE ACCESS INMEMORY FULL	MY_SALES	14M	336M	1070 (27)	00:00:01

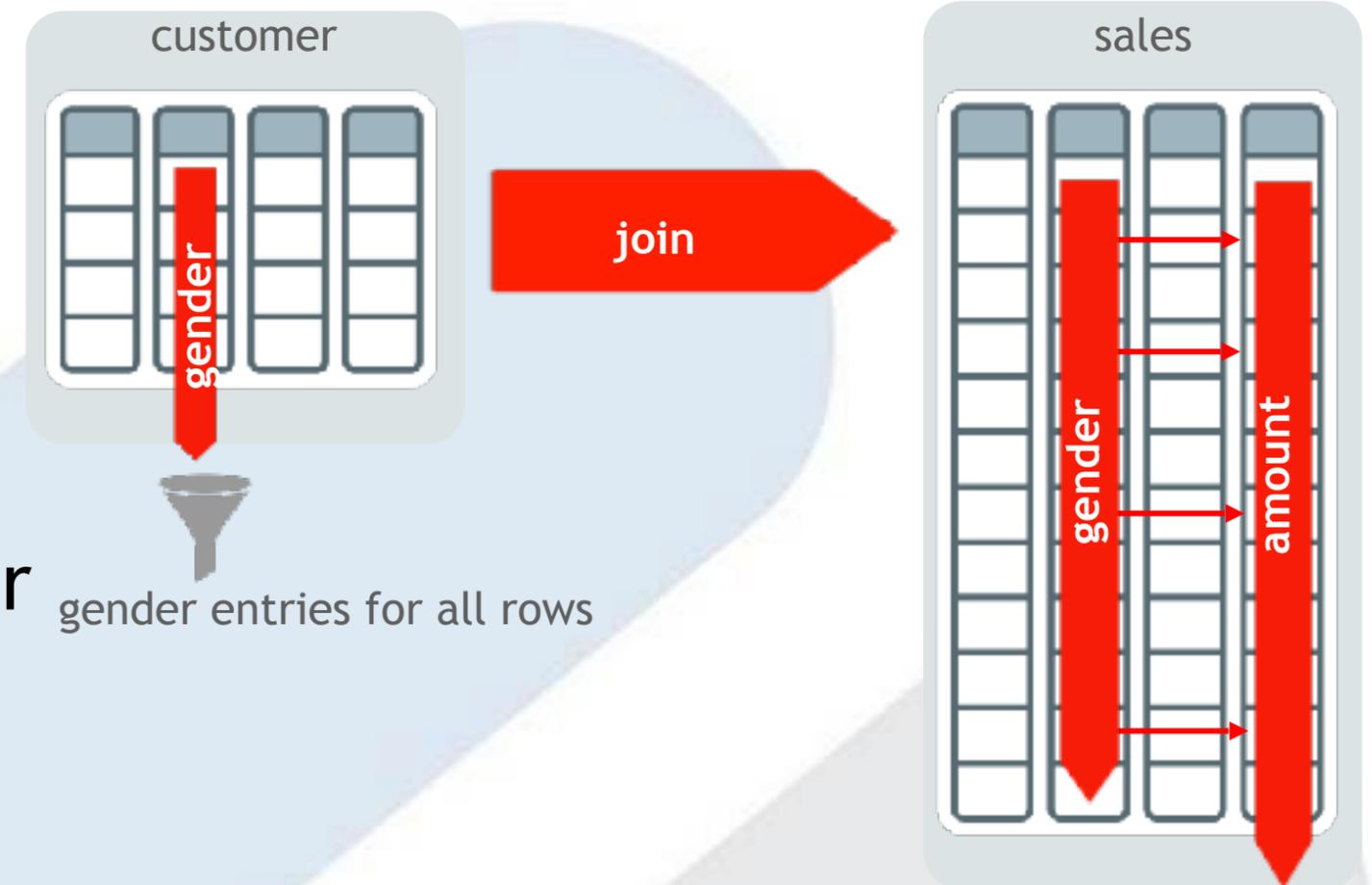
- With explicit NO_INMEMORY Hint**

3. Huge cost advantage due to IM

Id	Operation	Name	Rows	Bytes	Cost (%CPU)	Time
0	SELECT STATEMENT		1	24	19350 (1)	00:00:01
1	SORT GROUP BY		1	24		
2	TABLE ACCESS FULL	MY_SALES	14M	336M	19350 (1)	00:00:01

2. Switched to row storage SGA, Flash, or Disk

- Join tables on flexible criteria
- No special model required, e.g. star schema
- No indices required
- Speed up joins by use of bloom filters (not on all occasions)
- Small hash joins in PGA
- However data has to be decompressed for
 - bloom filter construction
 - hashing



Real-Time Analytics



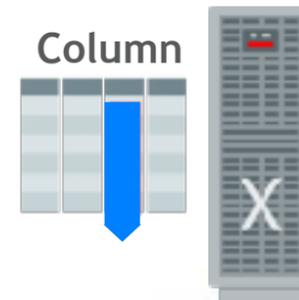
2X Faster Joins
5X Faster Expressions

Mixed Workload



Active Data
Guard Support

Massive Capacity



In-Memory on
Exadata Flash

Multi-model



Native support for
JSON Data type

Automation



Dynamic Data
Movement Between
Storage & Memory

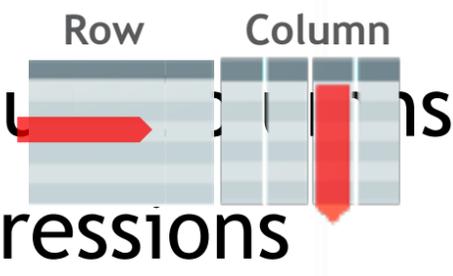
Real-Time Analytics

- Join Groups



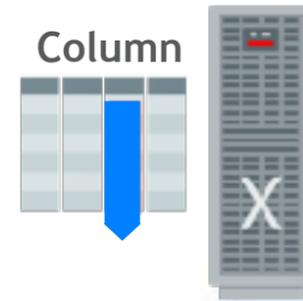
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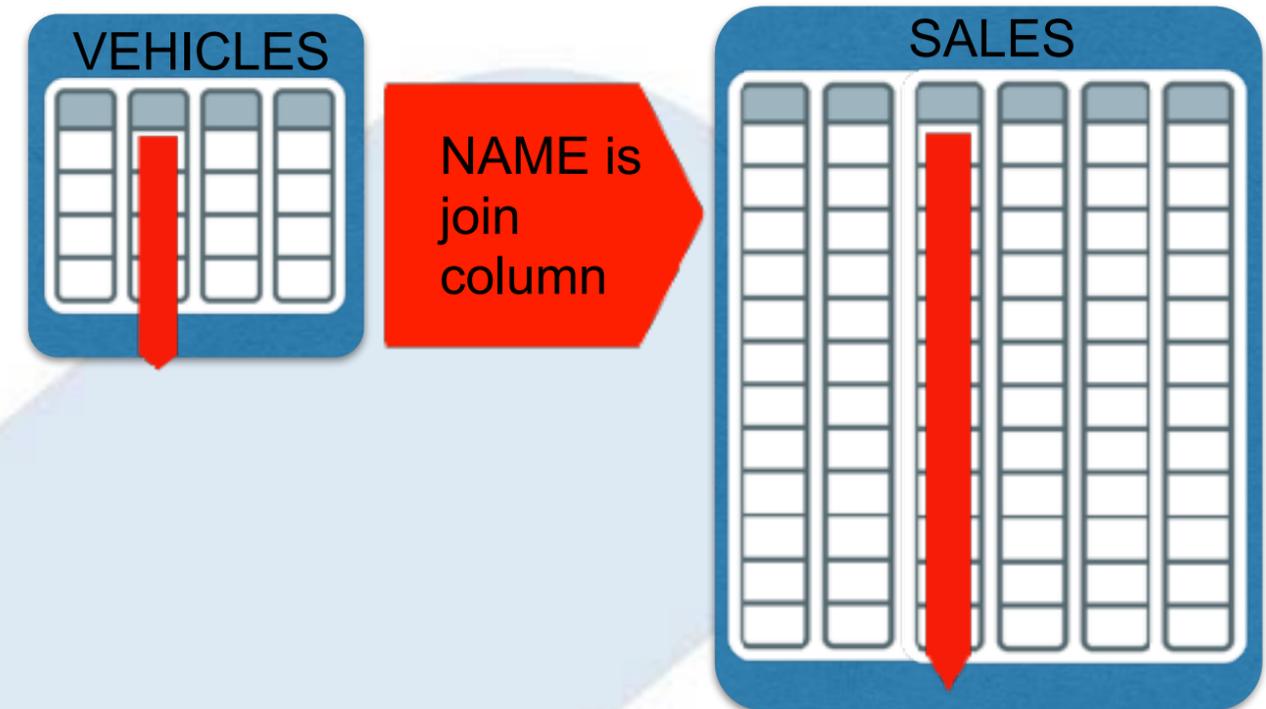


Dynamic Data
Movement Between
Storage & Memory

Join Groups: Faster Hash Joins

- Join columns in both tables are compressed using the same dictionary
- Joins occur on dictionary values rather than on data
 - Saves on decompression of data
 - Save on hashing the data

Example: Find sales price of each Vehicle



Local Dictionary at CU level

Column value list

BMW
Audi
BMW
Cadillac
BMW
Audi
Audi



Column CU	
Min: Audi	
Max: Cadillac	
VALUE	ID
Audi	0
BMW	1
Cadillac	2
1	
0	
1	
2	
1	
0	
0	

- Contiguous storage per column in an IMCU
- All CUs automatically store Min/Max values
- Multiple formats: depends on data and chosen compression level
- Most CUs have a “Dictionary”
 - Sorted list of distinct values in the CU
 - Column values replaced with dictionary IDs
- Additional compression of value list possible (e.g. Run Length Encoding, OZIP)

Same Global Dictionary used by Both Tables

Global Dictionary

NAME	ID
AUDI	0
BMW	1
CADILLAC	2
PORSCHE	3
TESLA	4
VW	5



- Global Dictionary created when 1st table is populated & used for join column in both tables

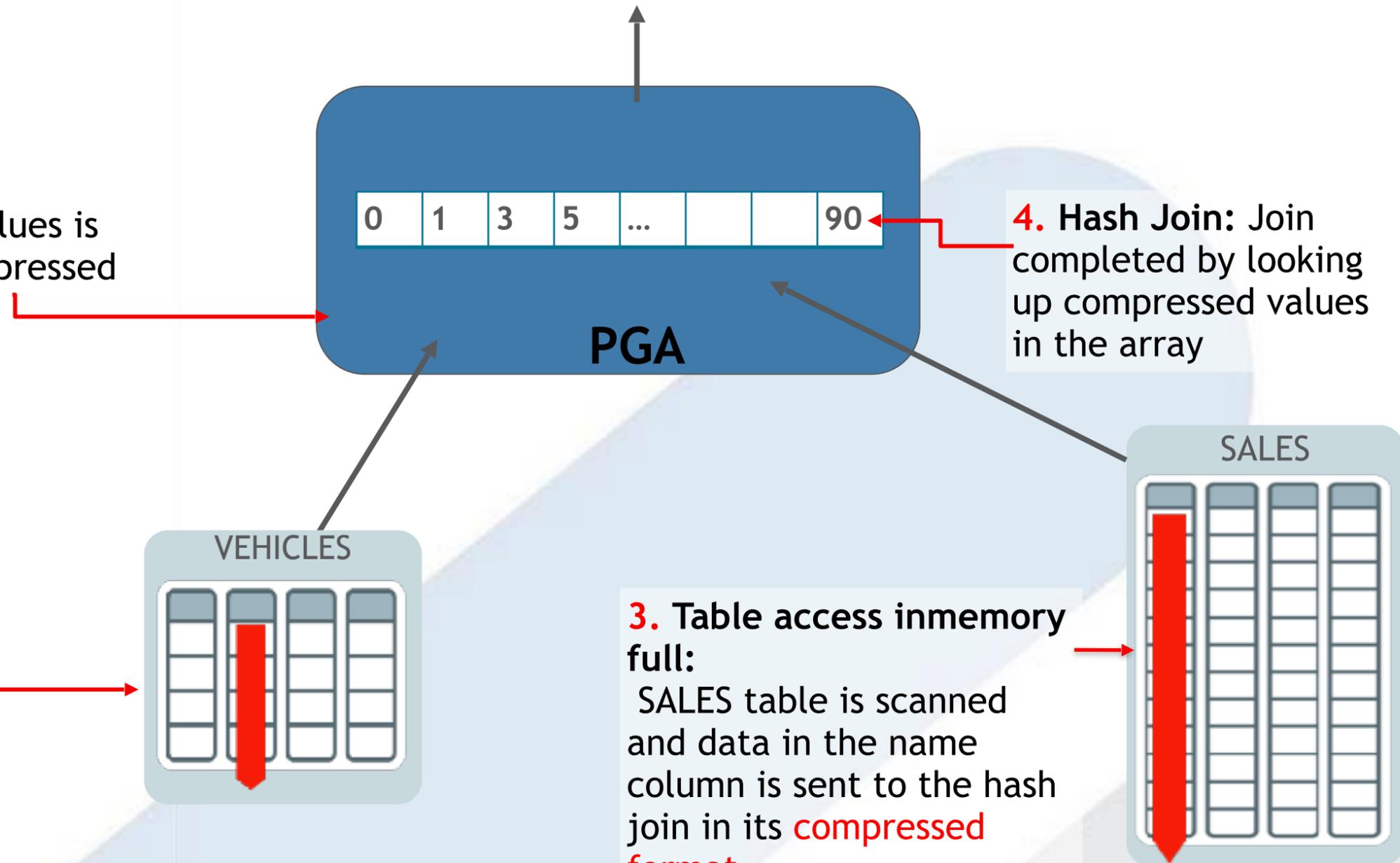
Hash Join With Join Group

2. Table Array of compressed values:
An array of distinct values is created from the compressed values

1. Table access inmemory full:
VEHICLES table is scanned and join column values sent to join in **compressed format**

3. Table access inmemory full:
SALES table is scanned and data in the name column is sent to the hash join in its **compressed format**

4. Hash Join: Join completed by looking up compressed values in the array



```
CREATE INMEMORY JOIN GROUP vehicle_ales_jg  
(VEHICLES (NAME) , SALES (NAME) ) ;
```

- Create Join Group
- Populate tables into column store
- Make sure that the global dictionary exists
- Use dictionary view `user_joininggroup` to monitor

```
SELECT o.object_name    table_name,  
       c.column_name    column_name,  
       gd.head_address  "GD Address"  
FROM   user_objects o,  
       user_tab_columns c,  
       v$im_segdict gd  
WHERE  gd.objn = o.object_id  
AND    o.object_name = c.table_name  
AND    gd.column_number = c.column_id;
```

Overview

General

SQL Text: SELECT --+ no_vector_transform monitor count("Custome ...)

Execution Started: Thu Dec 10, 2015 7:03:15 PM

Last Refresh Time: Thu Dec 10, 2015 7:03:20 PM

Execution ID: 16777232

User: VENDAVO

Fetch Calls: 1

Time & Wait Statistics

Duration: 5.0s

Database Time: 4.8s

PL/SQL & Java: 0s

Activity %: 100

Details

Plan Statistics | Plan | Activity | Metrics

Plan Hash Value: 2520514737 | Plan Note

Line ID	Operation	Name	Estim
0	SELECT STATEMENT		
1	SORT AGGREGATE		
2	HASH JOIN		
3	TABLE ACCESS FULL	V_PA_EXCHR_RT_JG	
4	PARTITION RANGE ALL		
5	TABLE ACCESS INMEMORY FULL	V_PA_TRANSACTIONS	

Other Plan Line Statistics

Build Size: 983K

Build Row Count: 380

Fan-out: 8

Slot Size: 123K

Total Build Partitions: 8

Total Cached Partitions: 8

Columnar Encodings Leveraged: 1

- Virtual Columns
 - do represent expressions involving columns in a table
 - are computed based on values within a record of a table
 - are not represented on disk
 - can be queried like any regular columns

- Syntax:

```
Alter table Lineorder
  Add (net_price AS
      ( l_extendedprice *
        (1 - l_discount) ) ) ;
```

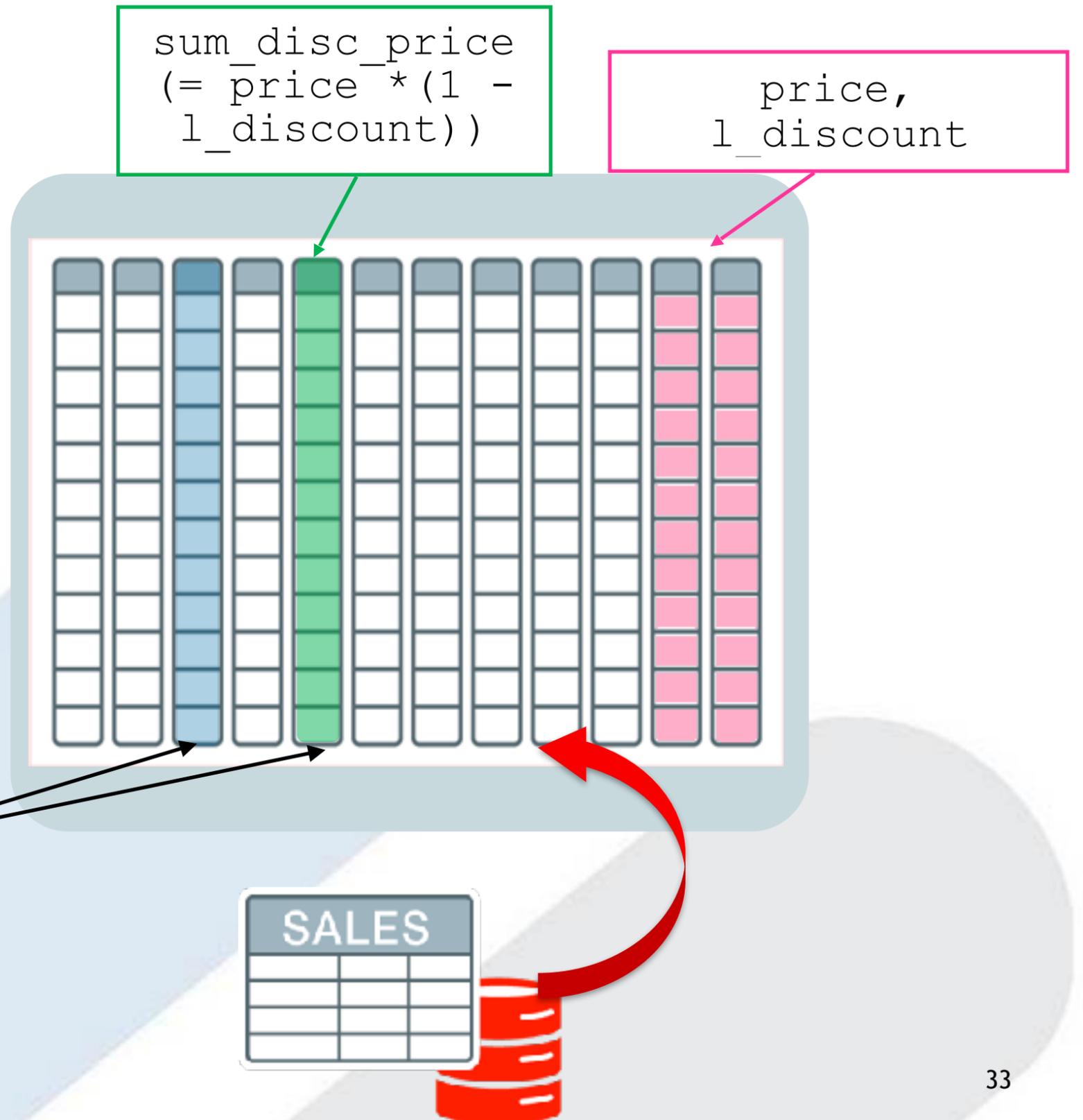
- Evaluating expressions repeatedly can be computationally expensive
- Repeated expressions are typical for analytic queries

```
select  l_returnflag, l_linestatus, sum(l_quantity) as sum_qty,  
        sum(l_extendedprice) as sum_base_price,  
        sum(l_extendedprice * (1 - l_discount)) as sum_disc_price,  
        sum(l_extendedprice * (1 - l_discount) * (1 + l_tax)) as charge,  
        avg(l_quantity) as avg_qty, avg(l_extendedprice) as avg_price,  
        avg(l_discount) as avg_disc,  
        count(*) as count_order  
From    lineorder  
Where   l_shipdate <= to_date ('1998-12-01', 'YYYY-MM-DD') - 90  
group by l_returnflag,  
         l_linestatus  
order by l_returnflag, l_linestatus;
```

How IM Virtual Columns Work

- Exist only in columnar store
- All performance features of In-Memory option are supported
 - IMCU pruning through storage indexes
 - SIMD
 - ...

```
Select sum_disc_price
From Sales
Where region = 'CA';
```



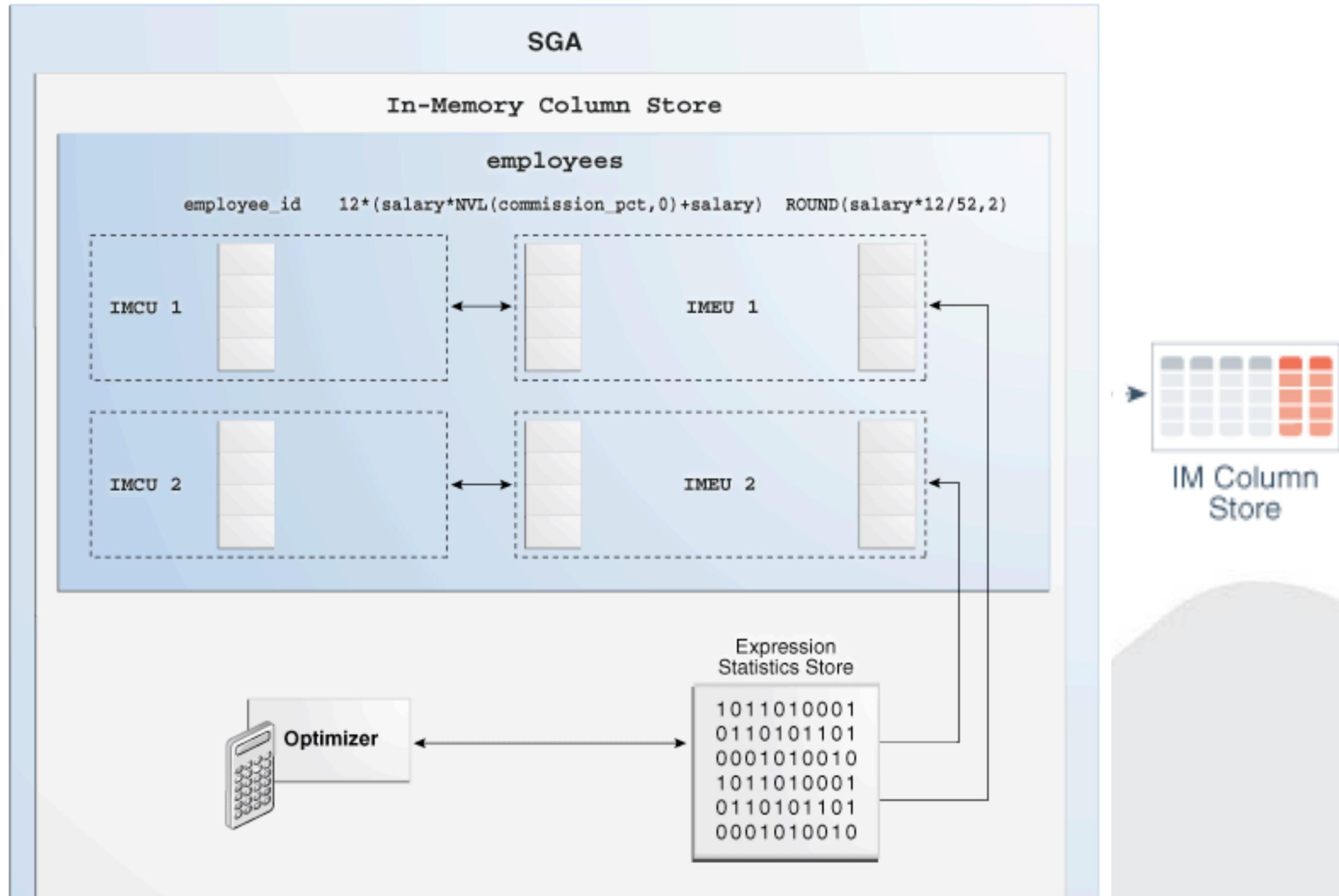
- Often used expressions that are automatically captured
- Capturing based on the Expression Statistics Store
- The database determines 'hot' expressions autonomously
- Top 50 expressions are added as virtual IM columns
- Populated automatically during next (re-)population of columnar store

Oracle SQL Developer : View SYS.ALL_EXPRESSION_STATISTICS@ssb_m1222

COLUMN NAME	DATA TYPE	NULLABLE	DATA DEFAULT	COLUMN ID	COMMENTS	INSERTABLE	UPDATABLE	DELETABLE
1 OWNER	VARCHAR2(128)	No	(null)	1	Owner of the table	NO	NO	NO
2 TABLE_NAME	VARCHAR2(128)	No	(null)	2	Name of the table	NO	NO	NO
3 EXPRESSION_ID	NUMBER	Yes	(null)	3	Expression ID of the current expression	NO	NO	NO
4 SNAPSHOT	VARCHAR2(10)	Yes	(null)	4	Type of Snapshot (cumulative or latest) for the expression	NO	NO	NO
5 EVALUATION_COUNT	NUMBER	Yes	(null)	5	Number of times the expressions has been evaluated	NO	NO	NO
6 FIXED_COST	NUMBER	No	(null)	6	Optimizer Fixed Cost of evaluating the expression	NO	NO	NO
7 DYNAMIC_COST	NUMBER	Yes	(null)	7	Optimizer Dynamic Cost of evaluating the expression	NO	NO	NO
8 EXPRESSION_TEXT	VARCHAR2(4000)	No	(null)	8	Text of the expression	NO	NO	NO
9 CREATED	DATE	No	(null)	9	Time this expression is first evaluated	NO	NO	NO
10 LAST_MODIFIED	DATE	Yes	(null)	10	Time this expression is last evaluated	NO	NO	NO

In-Memory Expression Units

Use
IME_CAPTURE_EX



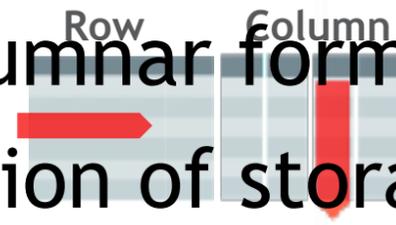
12.2 - Massive Storage Capacity

Real-Time Analytics

- In-Memory columnar format in flash cache
- Massive extension of storage capacity

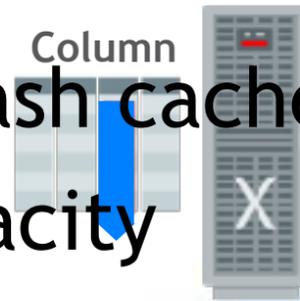
2X Faster Joins
5X Faster Expressions

Mixed Workload



Active Data
Guard Support

Massive Capacity



In-Memory on
Exadata Flash

Multi-model



Native support for
JSON Data type

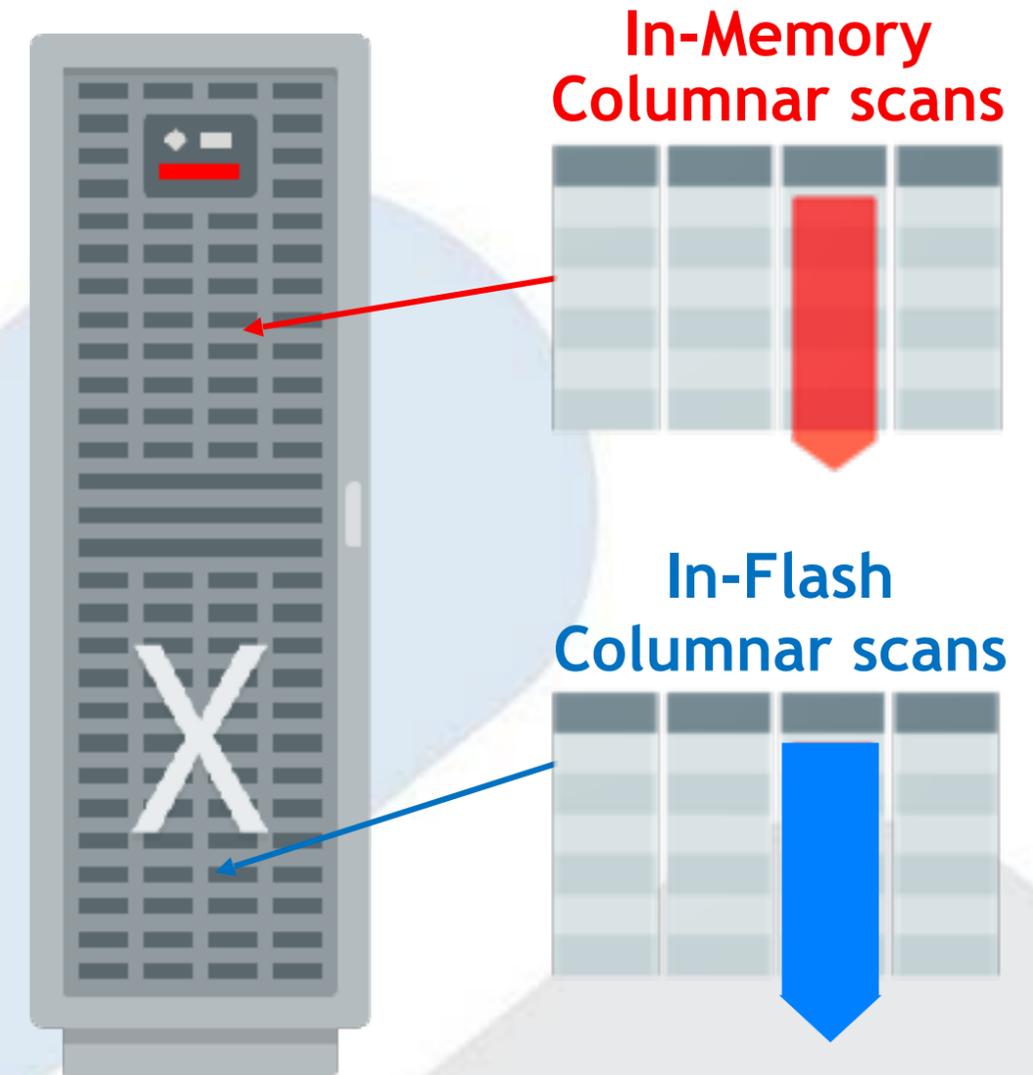
Automation



Dynamic Data
Movement Between
Storage & Memory

Use Of The Available Flash Cache

- New smart columnar flash cache format
- Available on Exadata
- Additional optimisations, e.g. evaluation of multiple column values in single vector instruction
- Optimiser spreads queries across DRAM & flash
- Extends available storage by factor 10
- Currently unique to Oracle In-Memory
- HCC objects are automatically populated into flash in IM-Columnar format



12.2 - Accelerate JSON Access

- Make use of JSON binary representation in the columnar store
- Support for multiple JSON functions since 12.1

Real-Time Analytics

Mixed Workload

Massive Capacity

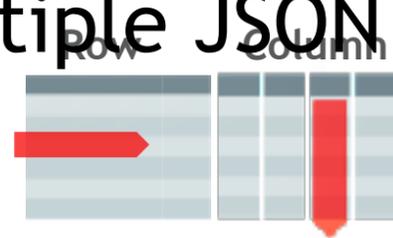
Multi-model

Automation

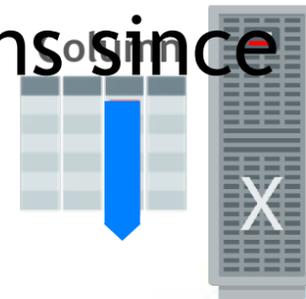
- json_table
- json_value

2X Faster Joins
5X Faster Expressions

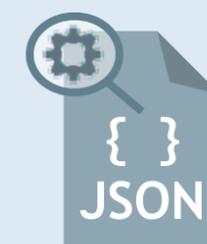
json_exists



Active Data Guard Support



In-Memory on Exadata Flash



Native support for JSON Data type



Dynamic Data Movement Between Storage & Memory

- In-Memory Expressions speed up JSON analytics further in IM 12.2

- Today's variety of analytical tools and tasks generate
 - more diverse workloads
 - require a very fast, flexible, agile, and robust query engine
- Oracle In-Memory boosts many of today's analytical workloads
- The new 12.2 features
 - In-memory Expressions
 - Join Groups
 - JSON optimisations
 - In-Memory Columnar Storage in Flashprovide unprecedented performance
- Big Data SQL extends its use and integrates it with HADOOP clusters





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