Data Loading & Migration for Database Management Systems



Speaker: Adam Dziedzic Advisor: Prof. Aaron Elmore



CSV vs. DBMS







- **►** BIG **RAW** DATA files
- Expensive raw analysis
- Fresh data = interesting data but unexplored
- Exponential growth

- ☑ ACID guarantees
- **☑** SQL interface
- ✓ Top query performance& convenient, fast analysis
- ☑ Linear complexity of loading

CSV vs. DBMS







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Only a small fraction of raw data is loaded to DBMSs [IDC 12]

How to accelearate data loading & migration?

1. Loading:

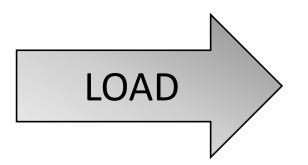
- a) CSV vs. DBMS
- b) General principles
- c) Thread/Process Level Parallelism
- d) Identify Bottlenecks
- e) Using Storage Devices
- f) Data level parallelism (SIMD)

2. Migration:

- a) Leverage Diverse Databases
- b) CSV vs. Binary format

Data Loading

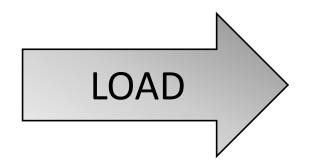






Data Loading







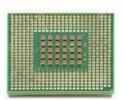
Read

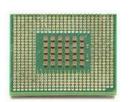
Process

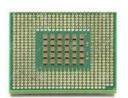
Write

HDD









RAID-0 (24 disks)





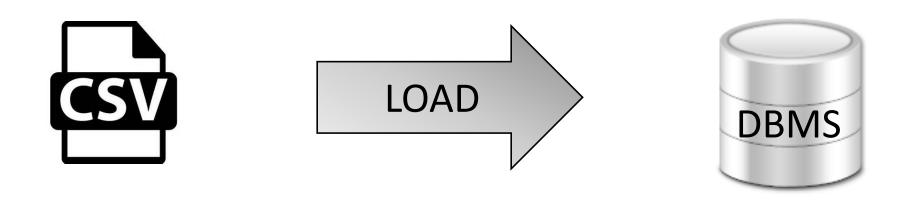




Data Loading: general principles

- 1. DON'T USE SLOW INSERT INTO or prepared statements.
- 2. **USE BULK LOADING <u>COPY INTO</u>** in PostgreSQL, MonetDB, Vertica, parallel load via external tables in Oracle, bulk insert in SQLServer.
- 3. Ask you data provider for:
 - a) Dividing the raw data into many files (few GBs) and load them in parallel the fastest method to achieve good performance;
 - b) Data in a binary format and a fast parser to process them;
- 4. Clean your data:
 - a) PostgreSQL ABORTS loading when data is ill-formatted;
 - b) Other databases store ill-formatted rows in *reject tables* or *error files* but all the well-formatted rows are loaded.

BULK Data Loading



How long does it take to load data into PostgreSQL?

Data Loading – current state of the art

TPC-H data 10GB, 2 sockets, 8 cores/socket, 2GHz, 20MB L3, 64GB RAM, from HDD->DAS, RHEL 6

Database	Loading time (seconds)
PostgreSQL	395.39
DBMS-B (commercial column-store)	234.94
MonetDB (open-source column-store)	149.11
DBMS-A (commercial row-store)	137.59

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DBMS-A (commercial row-store)	137.59
PARALLEL loading to PostgreSQL (with a single reader)	104.37

How to improve the performance of data loading?

How to accelearate data loading and migration?

1. Loading:

- a) CSV vs. DBMS
- b) General principles
- c) Thread/Process Level Parallelism
- d) Identify Bottlenecks
- e) Using Storage Devices
- f) Data level parallelism (SIMD)

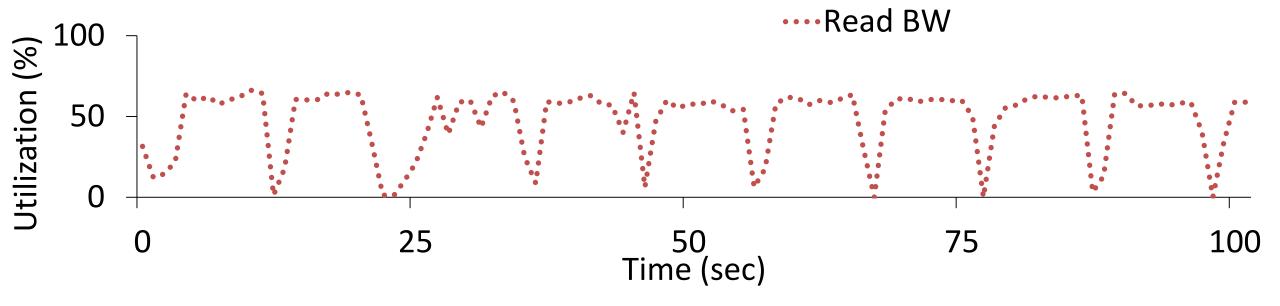
2. Migration:

- a) Leverage Diverse Databases
- b) CSV vs. Binary Migration

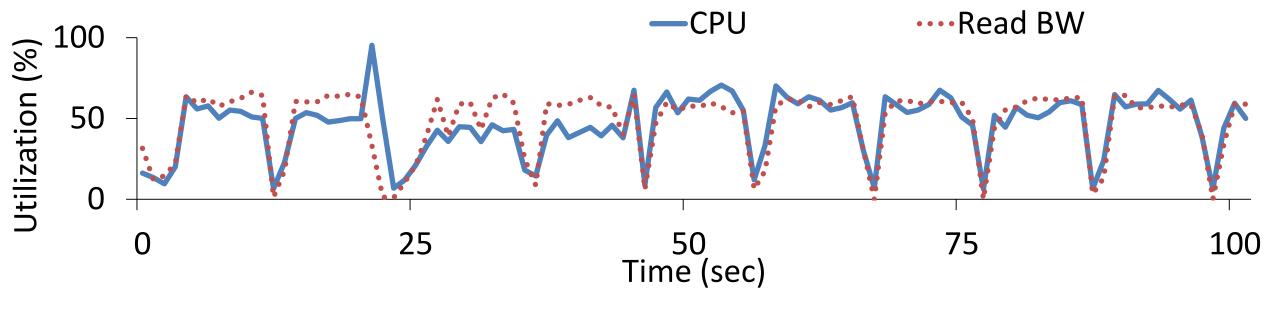
Resource utilization during Data Loading

LOAD: Read Process Write

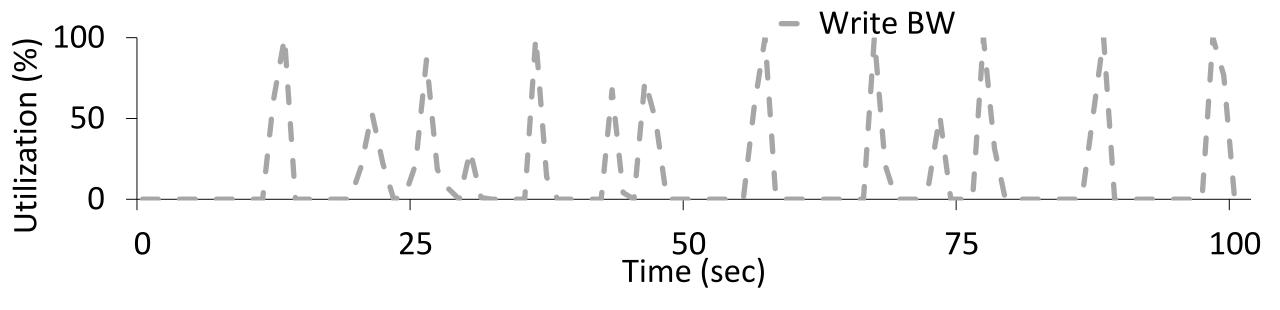
Read bandwidth underutilized



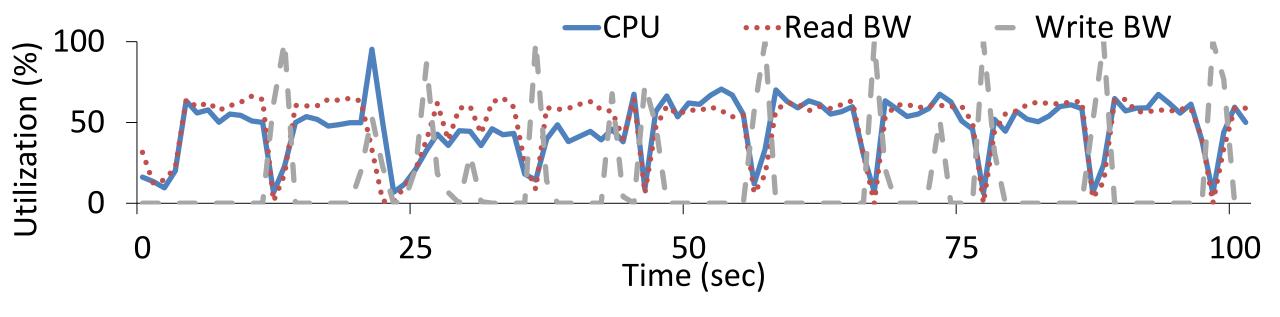
Low CPU utilization



Write bandwidth underutilized

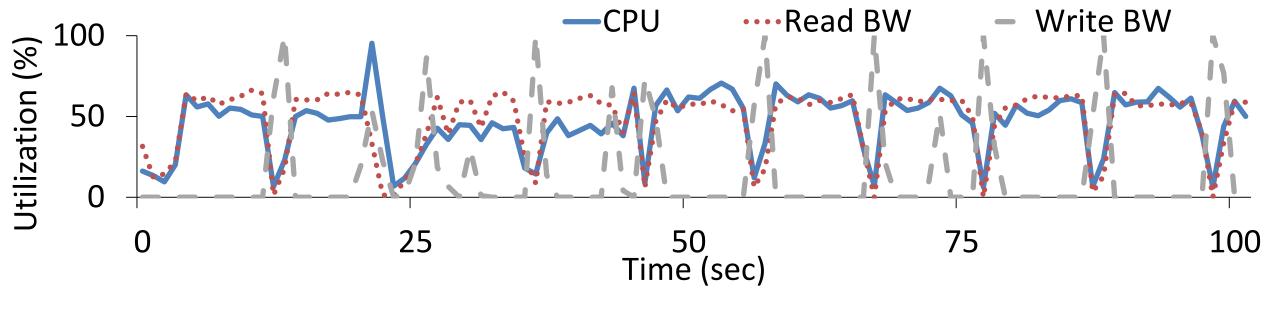


Alternating cycles in resource utilization



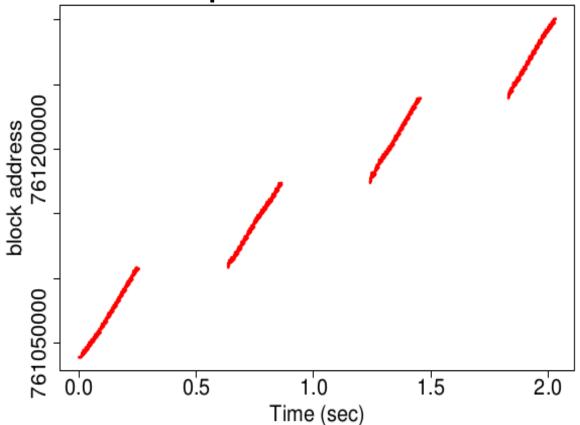
Unable to saturate resources





What causes the read bandwidth underutilization?

MonetDB Sequential reader



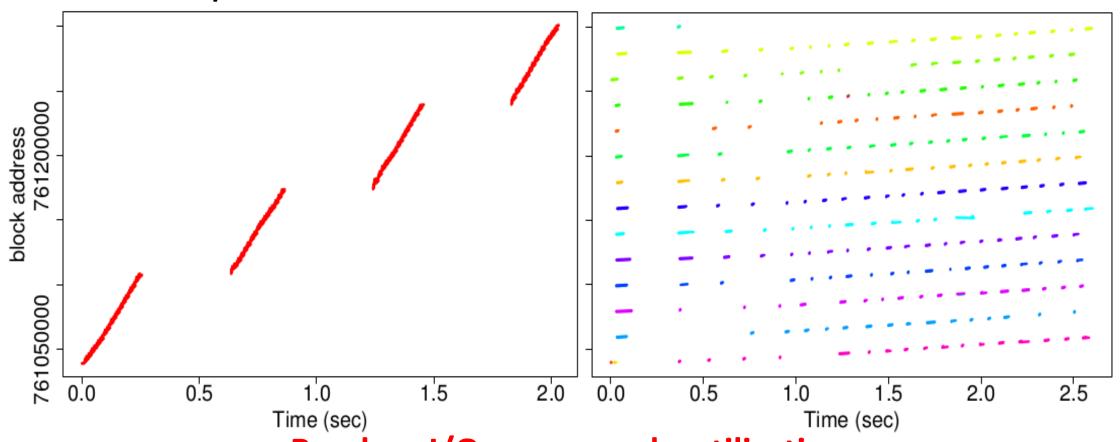
Orders TPC-H 10 GB HDD -> DAS, 16 threads, iosnoop

Read patterns

Read patterns

Reading orders file from TPC-H 10 GB; from HDD to DAS, 16 threads

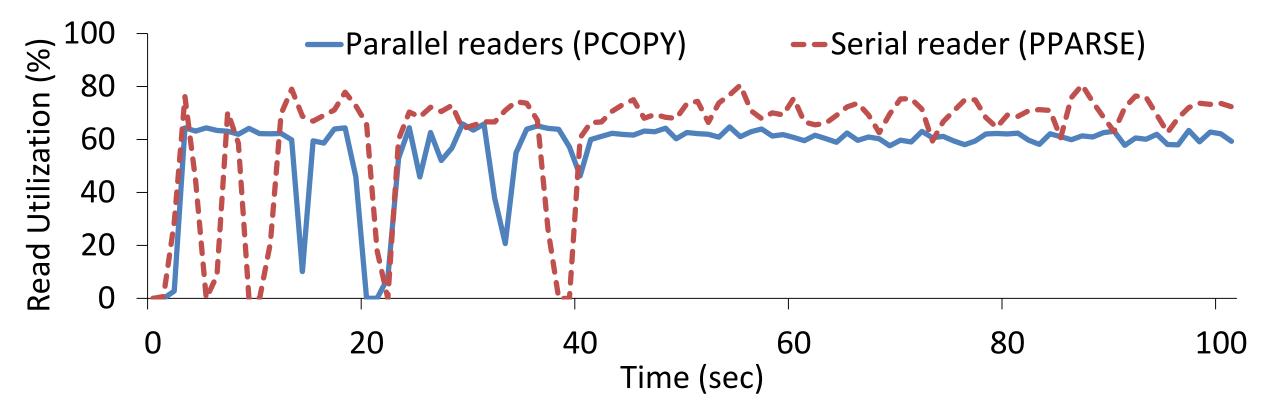
MonetDB Sequential reader DBMS-A
Parallel readers



Random I/O causes underutilization

Storage aware PostgreSQL: serial vs. parallel reader

TPC-H 10 GB from HDD to DAS; loading times: PCOPY 111 sec vs. PPARSE 104 sec



Serial reader improves read utilization # readers depends on input device

How to accelearate data loading and migration?

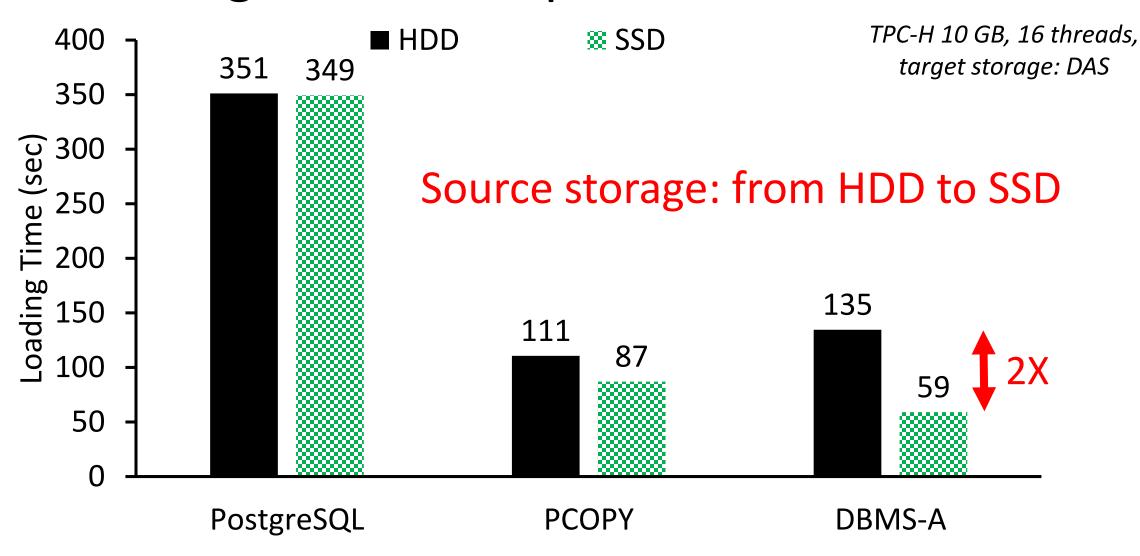
1. Loading:

- a) CSV vs. DBMS
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- e) Using **Storage Devices**
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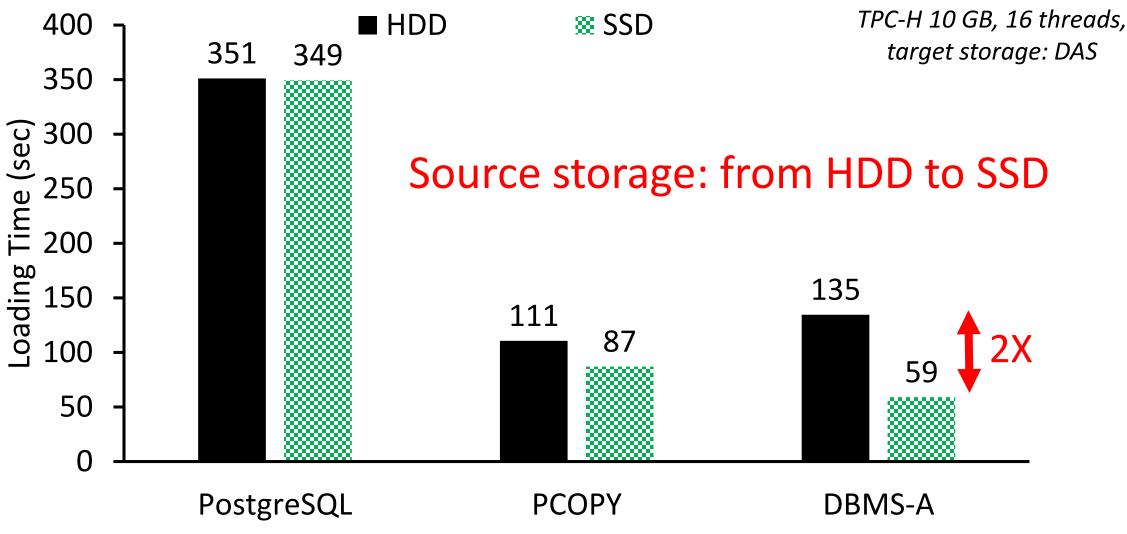
2. Migration:

- a) Leverage Diverse Databases
- b) CSV vs. Binary format

Storage device for parallel READERS

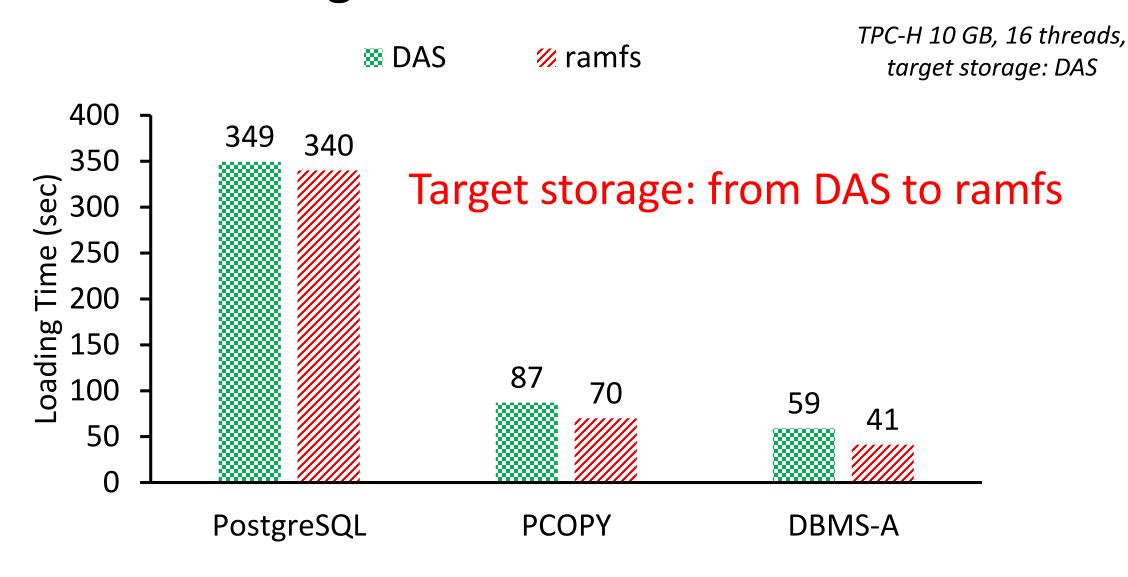


Storage device for parallel READERS



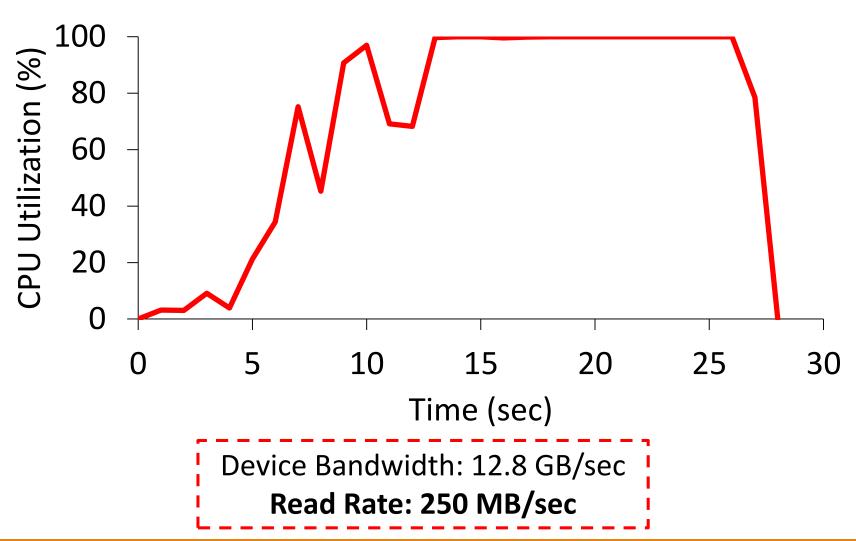
Single reader on HDD & Parallel readers on SSD

Storage device for WRITES

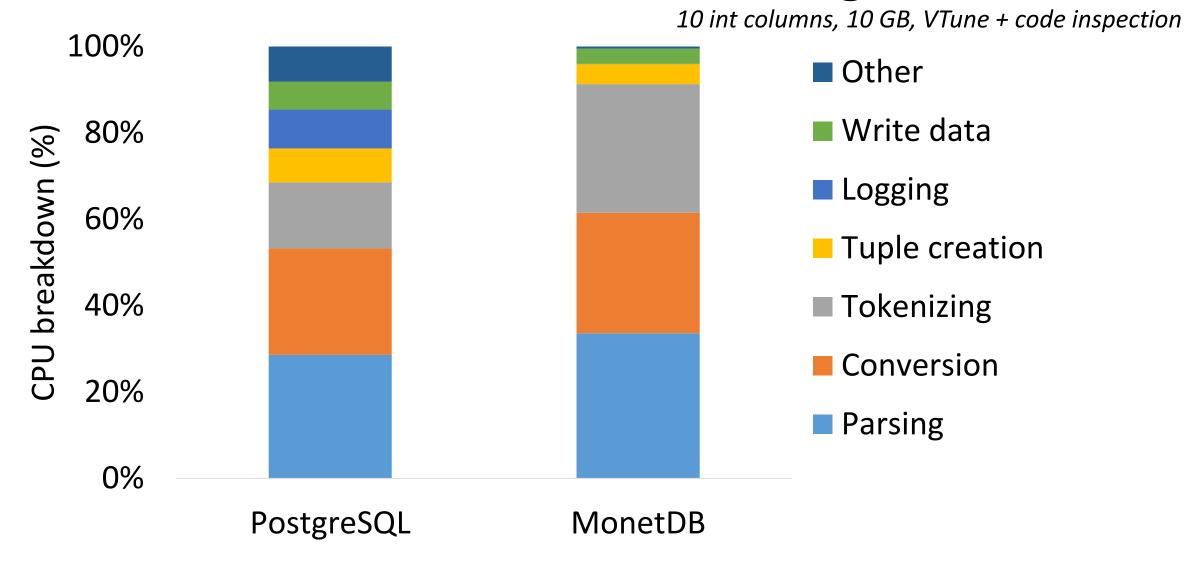


Best case storage scenario

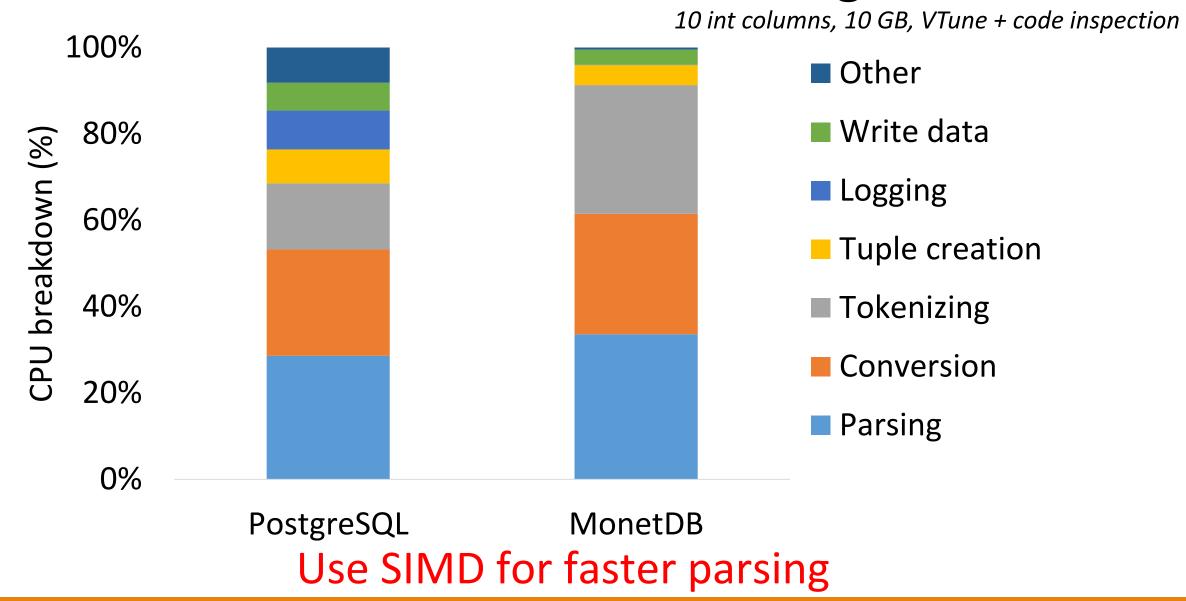
DBMS-A, TCP-H 10 GB, from ramfs to ramfs, 16 threads



Where does the CPU time go?



Where does the CPU time go?



How to accelearate data loading and migration?

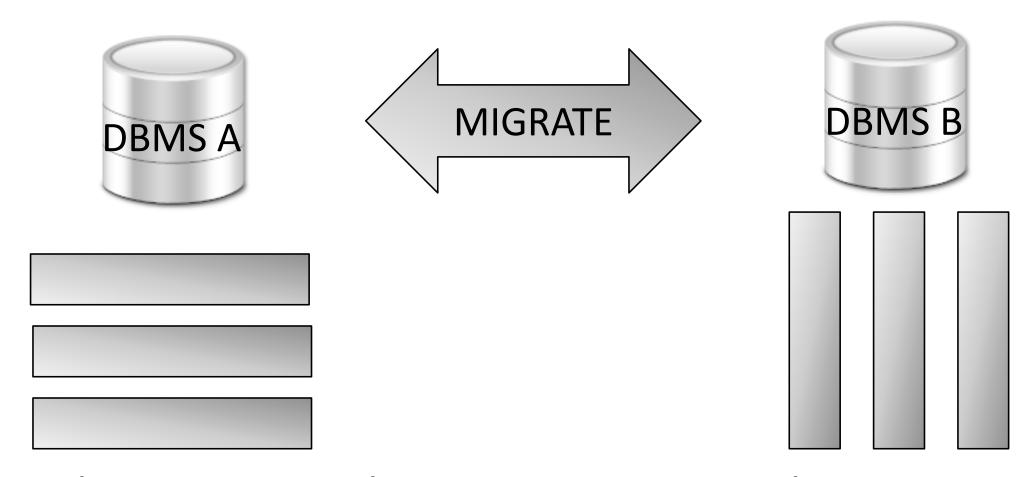
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Data Migration: leverge features of diverse databases



OLTP (row oriented)

Analytics (column oriented)

Current approach: CSV migration

CSV format

1,"Adam",6.00; 2,"Aaron",7.00

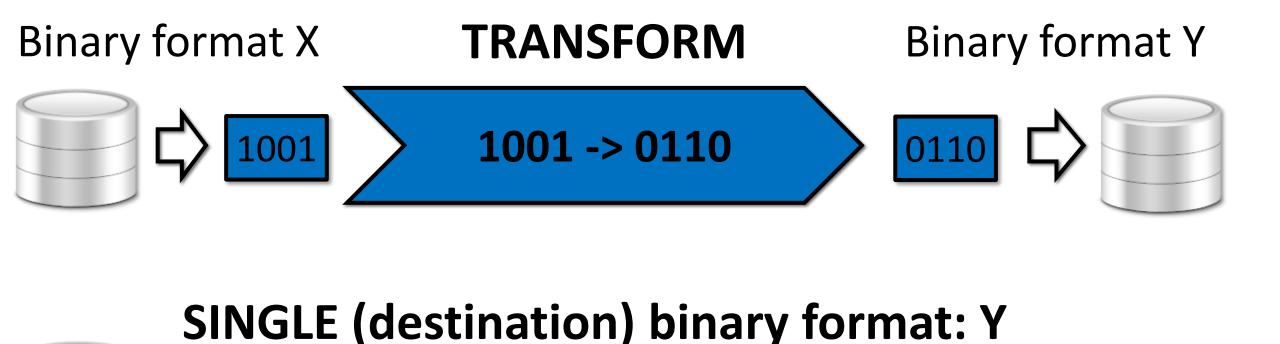


DBMS X

DBMS Y

Data already loaded to the source database

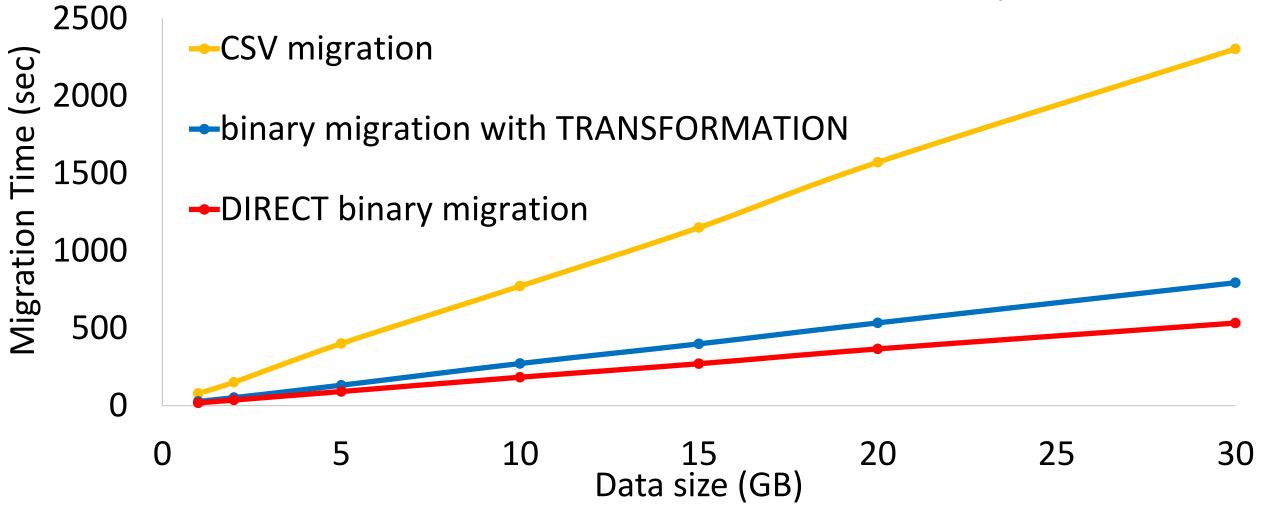
Our approach: binary migration



DBMS X DBMS Y

Data Migration from PostgreSQL to SciDB

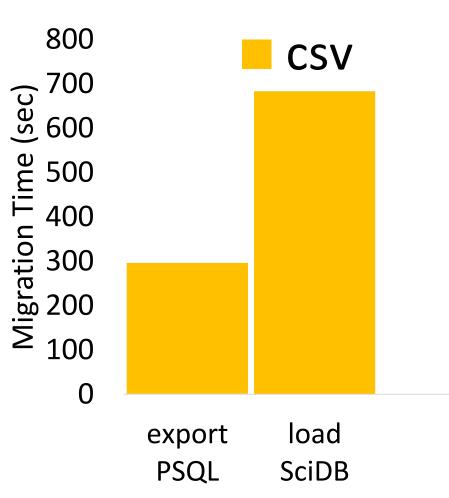
MIMIC II data - waveform(int, int, double)



TRANSFORMATION is 3X, DIRECT is 4X faster than CSV migration

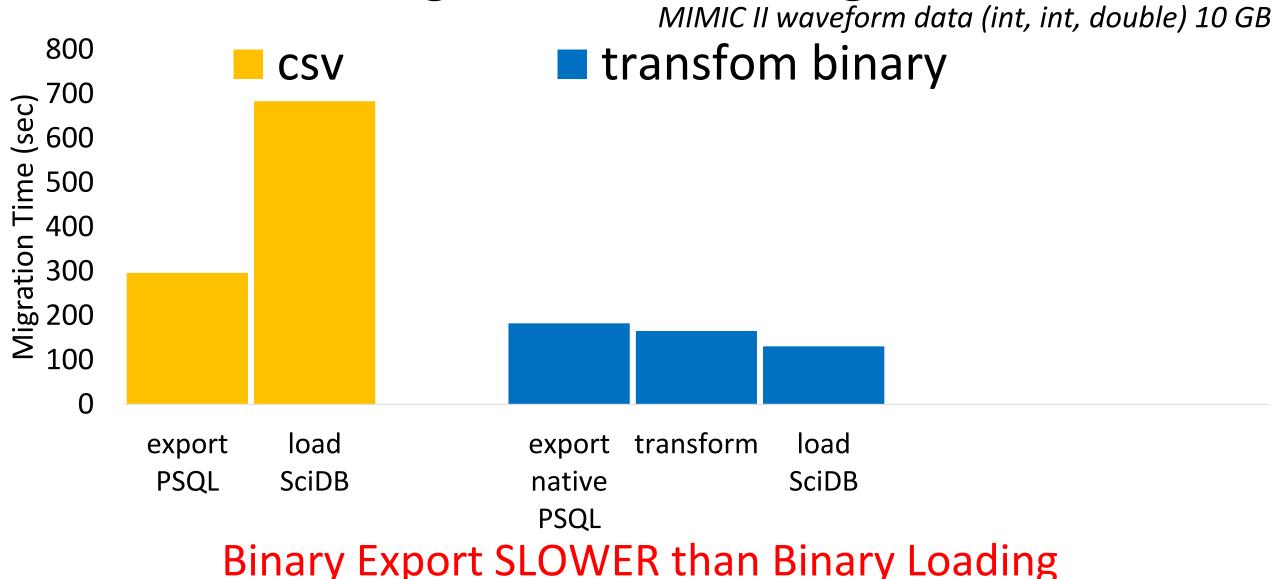
Breakdown: migration from PostgreSQL to SciDB

MIMIC II waveform data (int, int, double) 10 GB

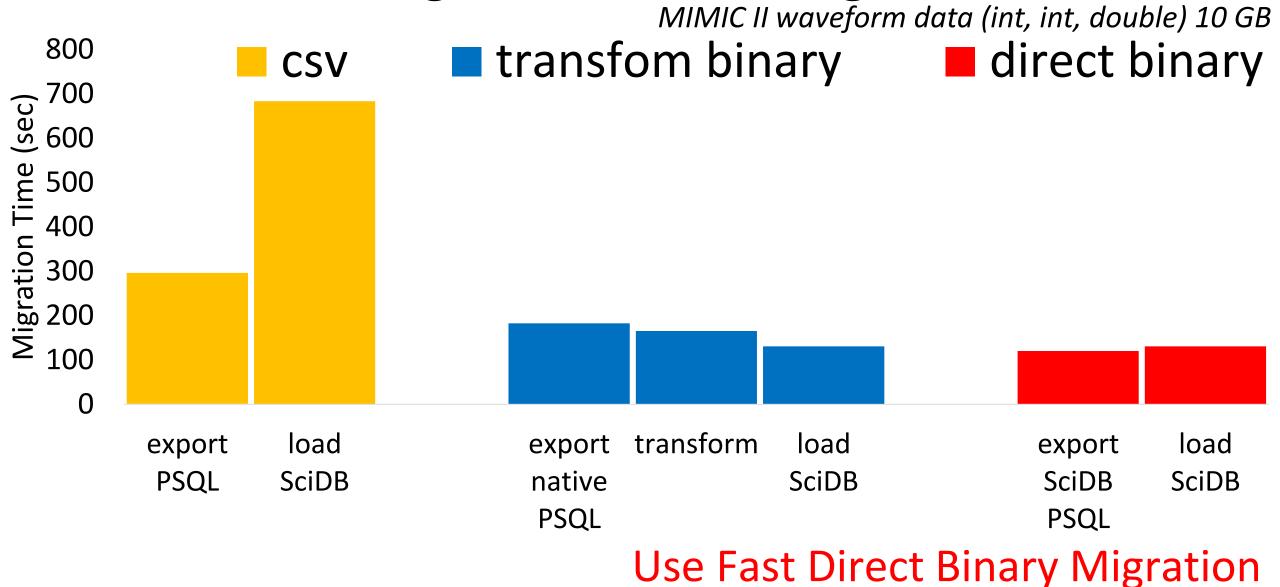


Slow CSV loading

Breakdown: migration from PostgreSQL to SciDB



Breakdown: migration from PostgreSQL to SciDB



Conclusions

- ☐ Parallelism is the key:
 - Ask the data provider for splitting the data during generation and BULK load the data files in parallel;
 - Tune your database and set up parallel loading;
 - Use SIMD (data level parallelism) for faster parsing.
- ☐ Optimize I/O to fully utilize CPU:
 - Input I/O path with single (HDD) or parallel reader (SSD);
 - Output I/O path to reduce pauses caused by data flushes.
- ☐ Binary data format for data transfers:
 - Select a single, concise, SIMD-friendly & binary format.

Thank you

Backup slides

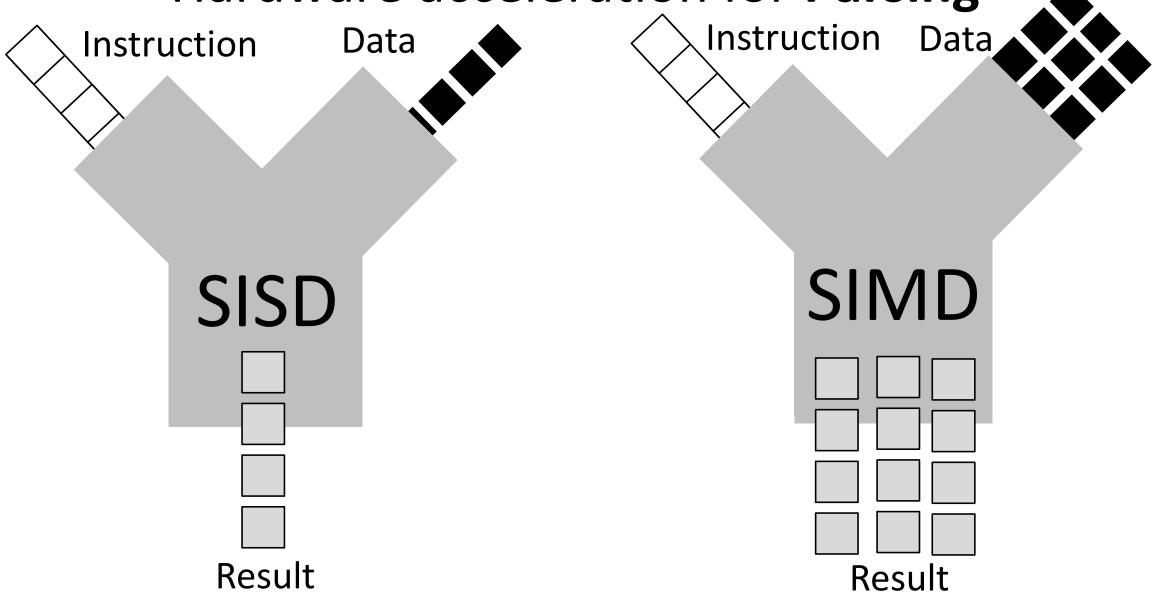
Faster loading, export & migration for PostgreSQL

- ☐ Load external loader for PostgreSQL:
 - Read data from many input files using many psql clients;
 - External parallel loader that is aware of storage devices.

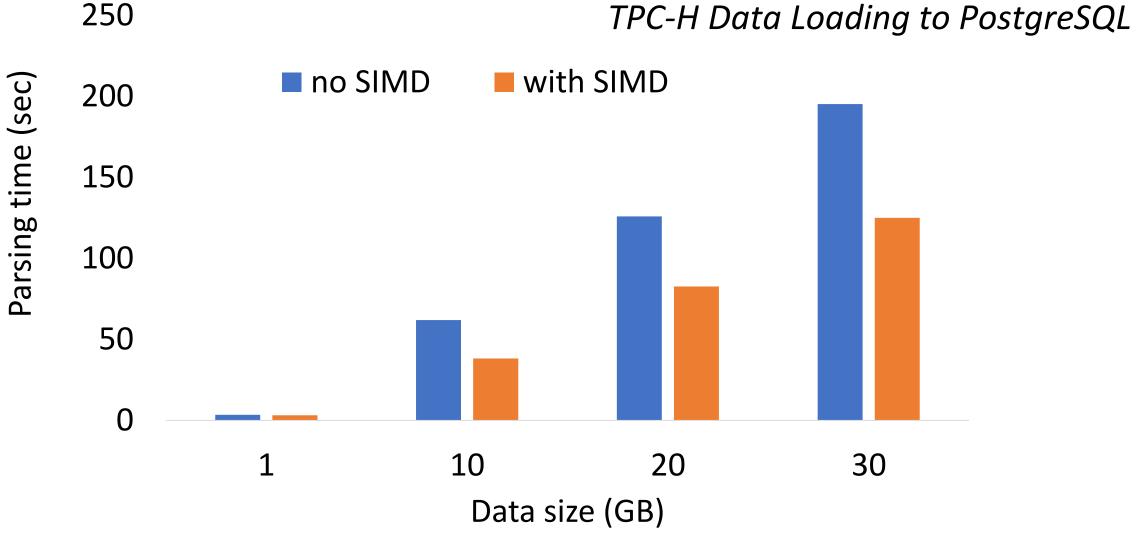
☐ Export:

- Both, parallel export in CSV format and export in a specific (non-postgres) binary format require changes to the PostgreSQL's source code.
- ☐ Migrate = export + load:
 - CSV: no changes to the database required (unless parallel);
 - Binary: either external transformation or internal changes.

Hardware acceleration for Parsing



SIMD: for parsing lines in input CSV file



SIMD gives 1.6X speedup for parsing lines

How to accelearate data loading, export and migration?

1. Loading:

- a) Intro: CSV vs. DBMS, and basic recommendations;
- b) Thread/process level parallelism;
- c) Identify the bottlenecks;
- d) Impact of storage devices;
- e) Data level parallelism (hardware accelerator: SIMD).

2. Export:

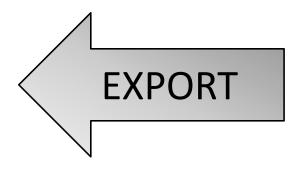
a) Extract database pages (collection of rows) to separate files with many processes.

3. Migration:

a) Single, concise, and binary data format for direct data migration.

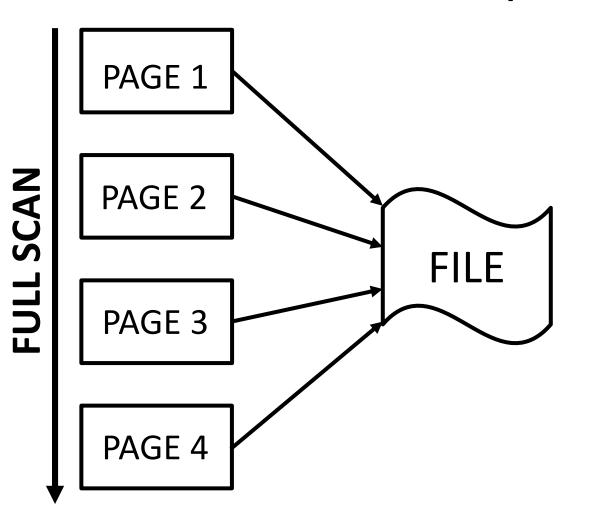
Data Export: don't get locked into a database vendor



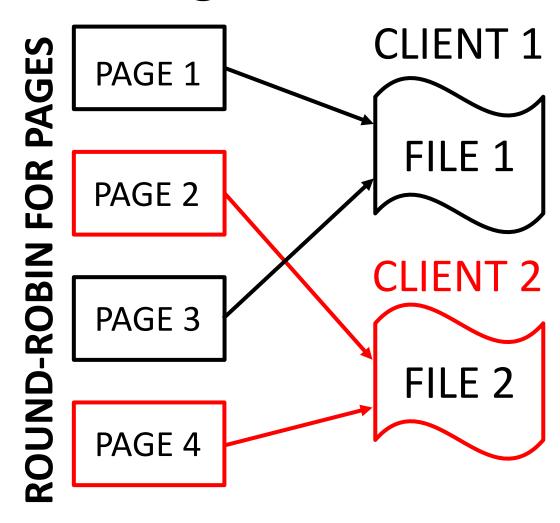




Parallel Export from PostgreSQL

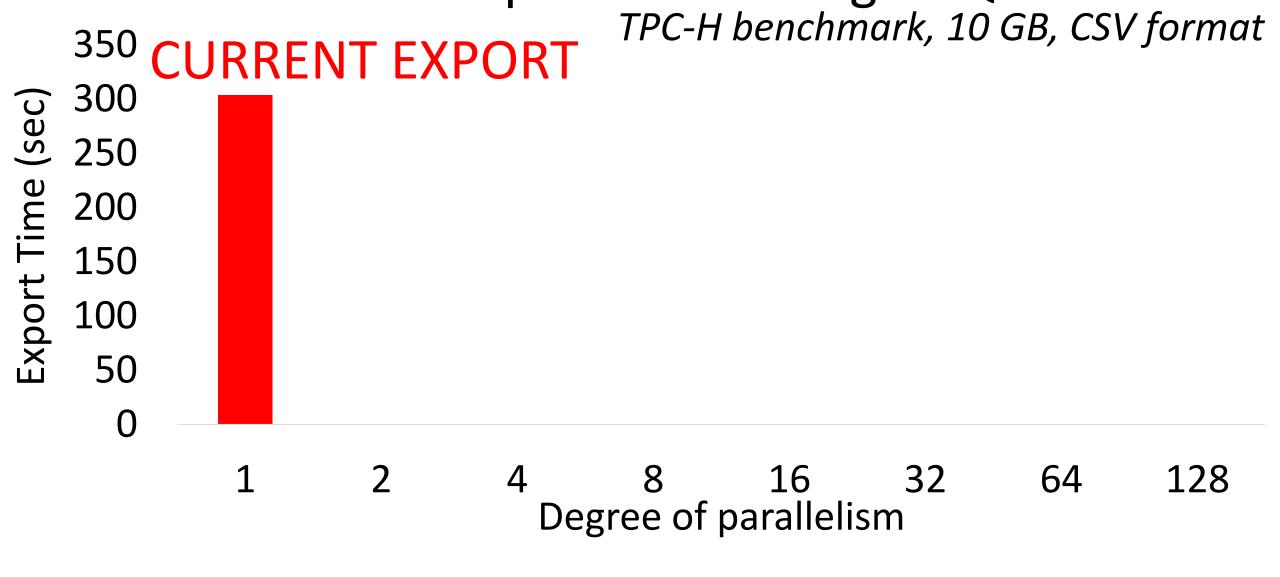


Current single-threaded export

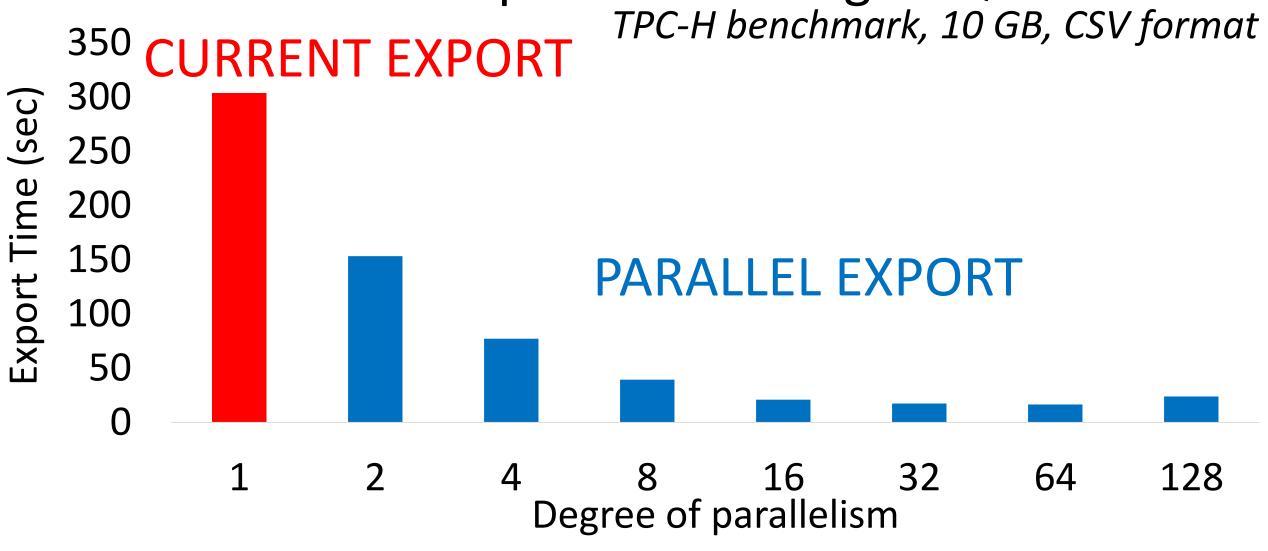


New parallel export

Parallel export from PostgreSQL

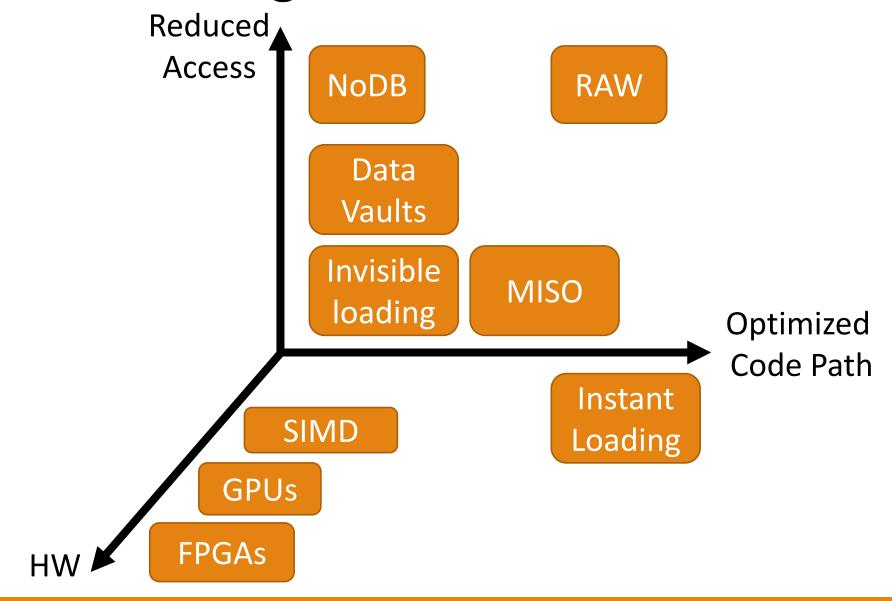


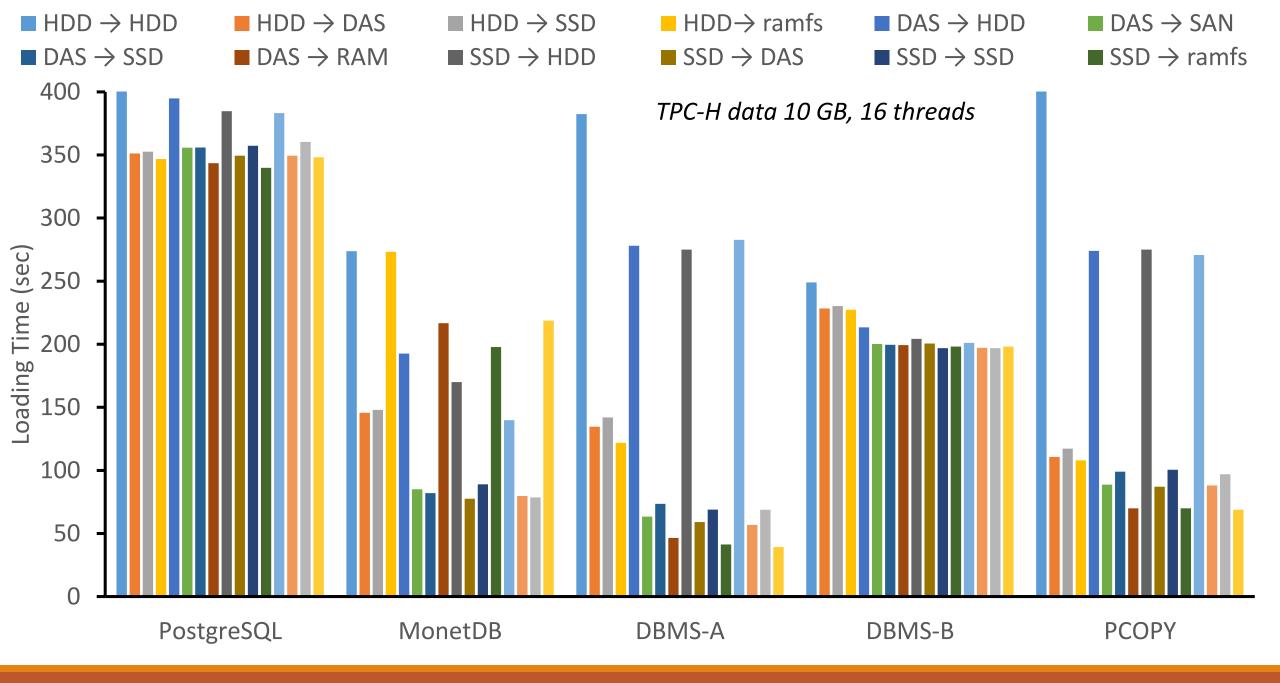
Parallel export from PostgreSQL



New Parallel export 20X faster than Current export

Reducing data loading overheads: related work





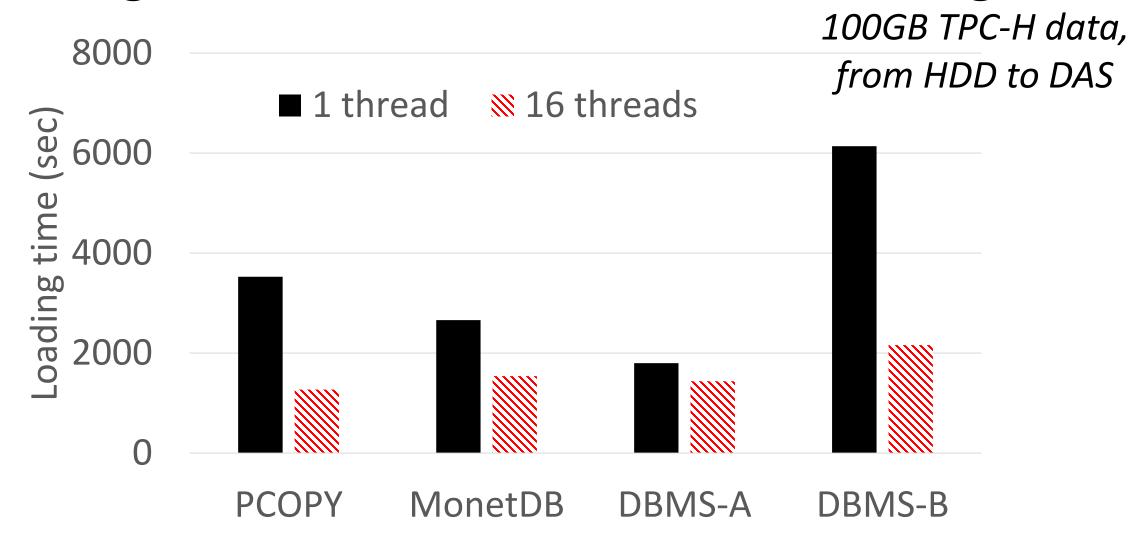
Experimental Setup

- ☐ Hardware
 - Dual socket 8 cores Intel(R) Xeon(R) CPU E5-2640, main memory: 64 GB
 - HDD: 4 x 500 GB 10k RPM SATA disks
 Max read/write throughput: 170 MB/s, 160 MB/s
 - DAS: 24 x 558 GB 10k RPM SATA disks (RAID-0)
 Max read/write throughput: 1100 MB/s, 330 MB/s
 - SSD: 3 x 200 GB, Max read/write throughput: 565 MB/s, 268 MB/s
- Software
 - Red Hat Enterprise Linux 6.6 (Santiago 64bit) with kernel version 2.6.32
 - Row-oriented DBMSs: PostgreSQL (v. 9.3.2), DBMS-A
 - Column-oriented DBMSs: MonetDB (v. 11.19.9), DBMS-B
 - PostgreSQL parallel external loaders: PCOPY and PPARSE
- Datasets
 - benchmark (TPC-H, TPC-C, SDSS) and real-world (Symantec, MIMIC-II)

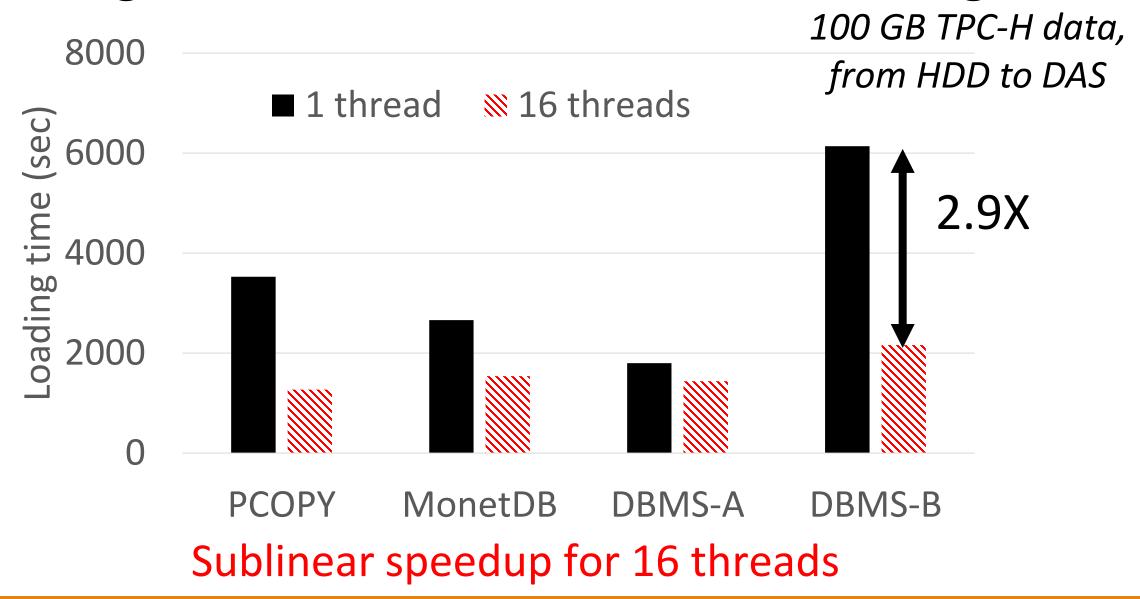
Storage devices

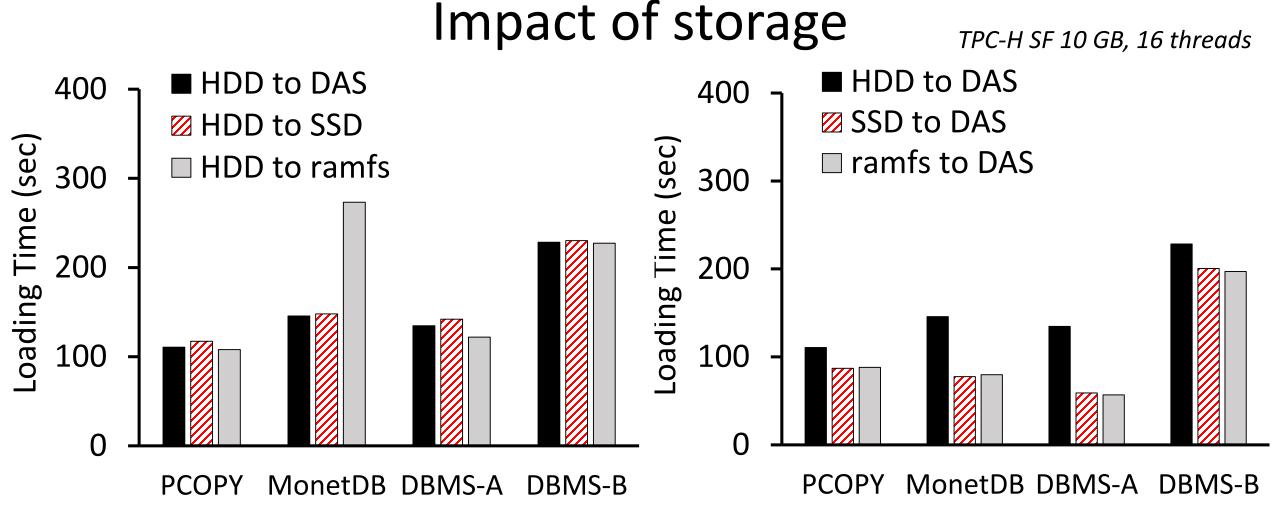
Name	Capacity	Configur ation	Read speed	Write speed	RPM
HDD	1.8 TB	4 x HDD (RAID-0)	170 MB/sec	160 MB/sec	7.5k
DAS	13 TB	24 x HDD (RAID-0)	1100 MB/sec	330 MB/sec	7.5k
SSD	550 GB	3 x SSD (RAID-0)	565 MB/sec	268 MB/sec	N/A
ramfs	64 GB	N/A	12.8 GB/sec	12.8 GB/sec	N/A

Single threaded vs. Parallel data loading



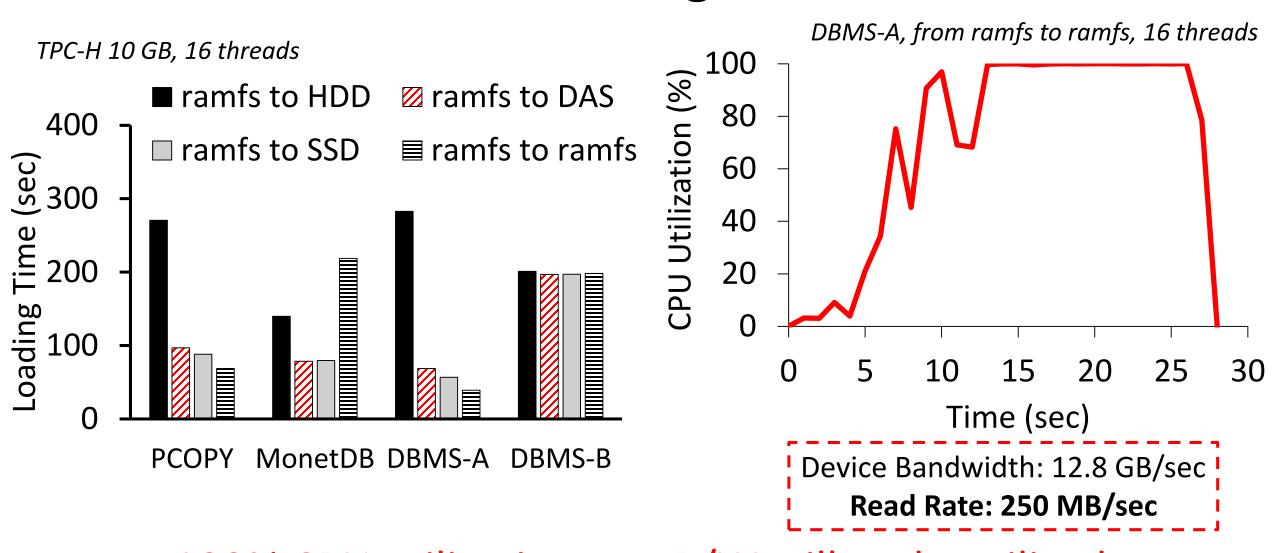
Single threaded vs. Parallel data loading





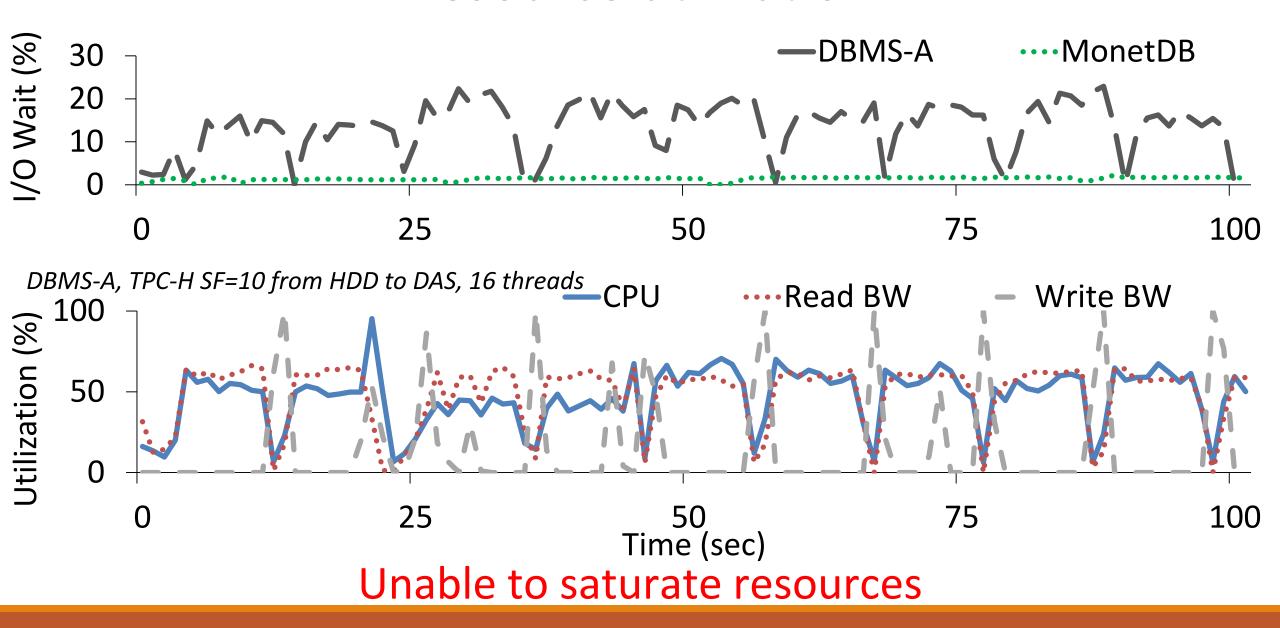
Slow source storage bottlenecks all systems Write bottleneck when source storage is fast

Best-case storage scenario



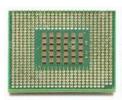
100% CPU utilization, yet B/W still underutilized

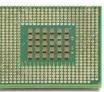
Resource utilization

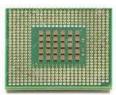


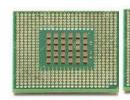
Resource utilization: I/O & CPU

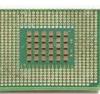
- ☐ Understand the lack of scalability under parallel loading
- ☐ Analyze the utilization of resources:
 - ☐ CPU
 - □ I/O
- ☐ What is the alternative view of the previous figures?

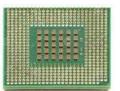














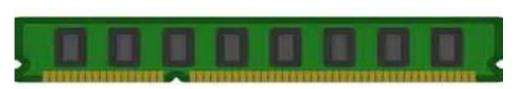
Impact of storage

- ☐ A typical DBMS setup underutilizes I/O bandwidth and CPU
- ☐ Problem: random I/O for parallel loading from HDD
- ☐ How different storage sub-systems affect data loading speed?

DBMS

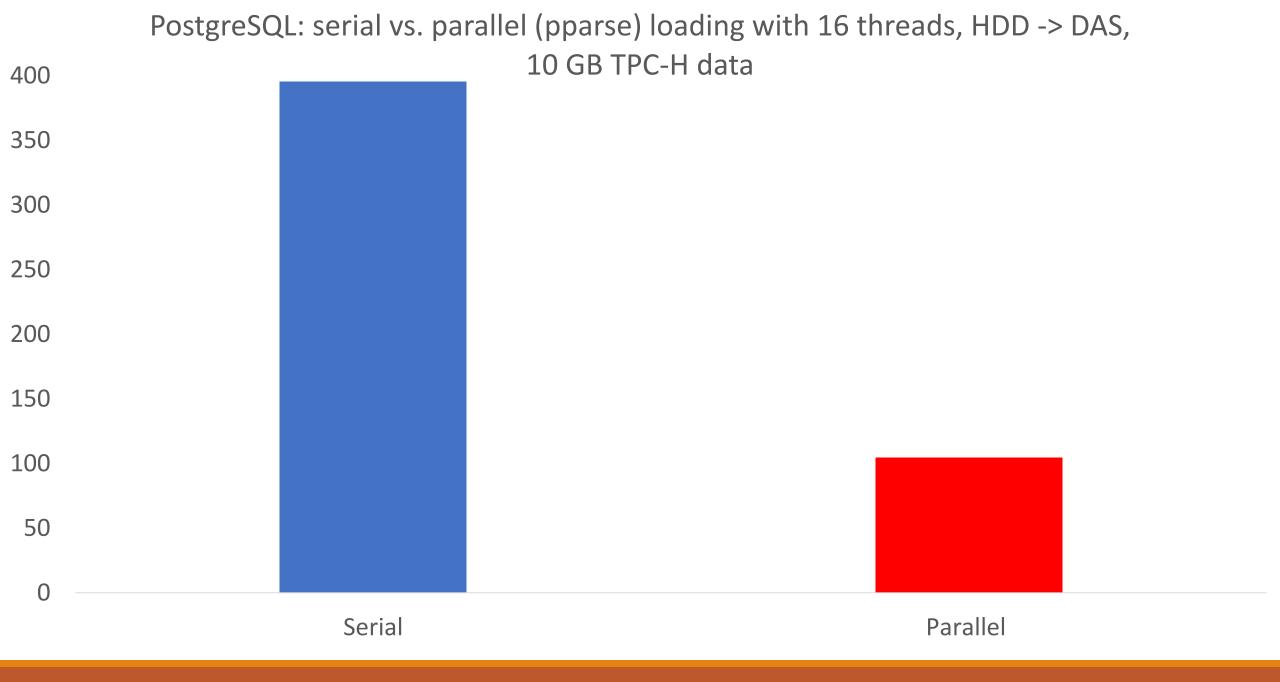






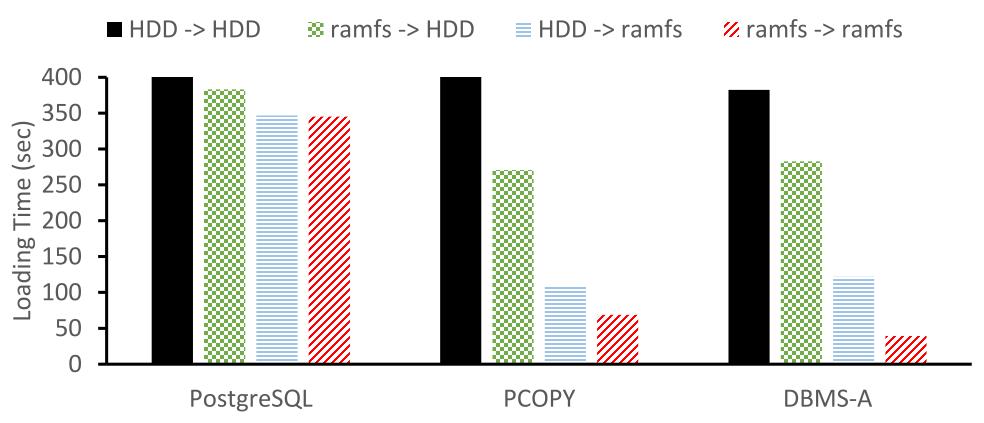






Impact of storage devices

TPC-H data 10 GB, 16 threads



Slow source storage bottlenecks all systems Write bottleneck when source storage is fast

High cost of Data Migration

2 nodes, each 4 cores 3.10 GHz, 4 MB L3, 16 GB RAM, SSD 250 GB, Ubuntu 14.04

METHOD	TIME (sec)		
From PostgreSQL to SciDB (MIMIC II data, 10 GB)			
CSV (common approach)	772		

From S-Store to SciDB (TPC-C data, 10 GB)				
CSV (common approach)	823			

High cost of Data Migration

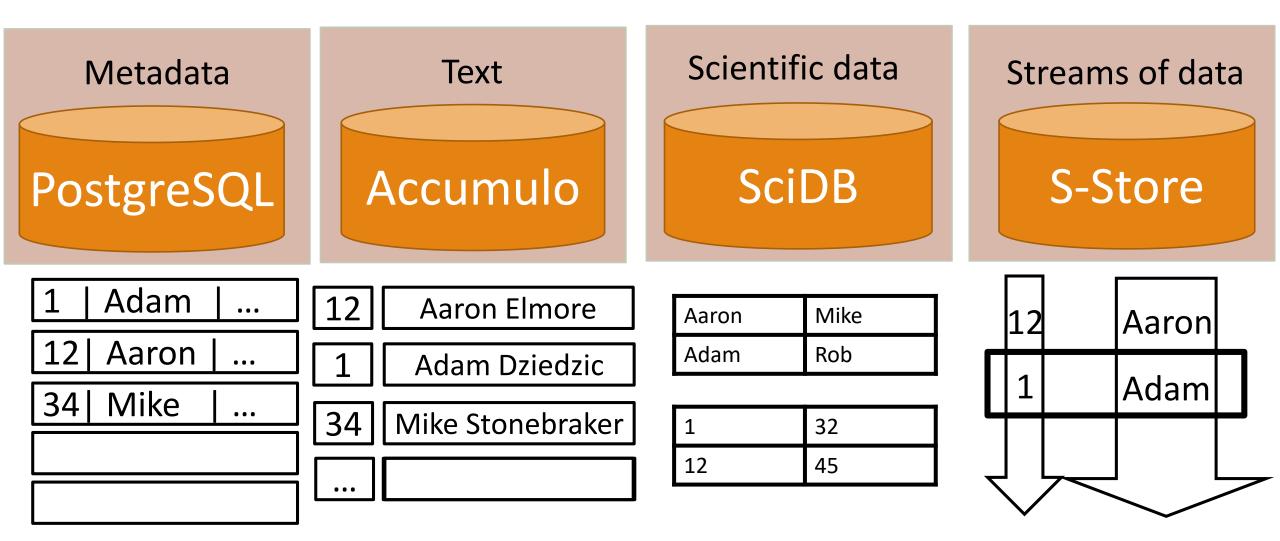
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METHOD	TIME (sec)	
From PostgreSQL to SciDB (MIMIC II data, 10 GB)		
CSV (common approach)	772	
Direct parallel binary migration with compression	75	

From S-Store to SciDB (TPC-C data, 10 GB)				
CSV (common approach)	823			
Parallel (16 X) direct binary migration	100			

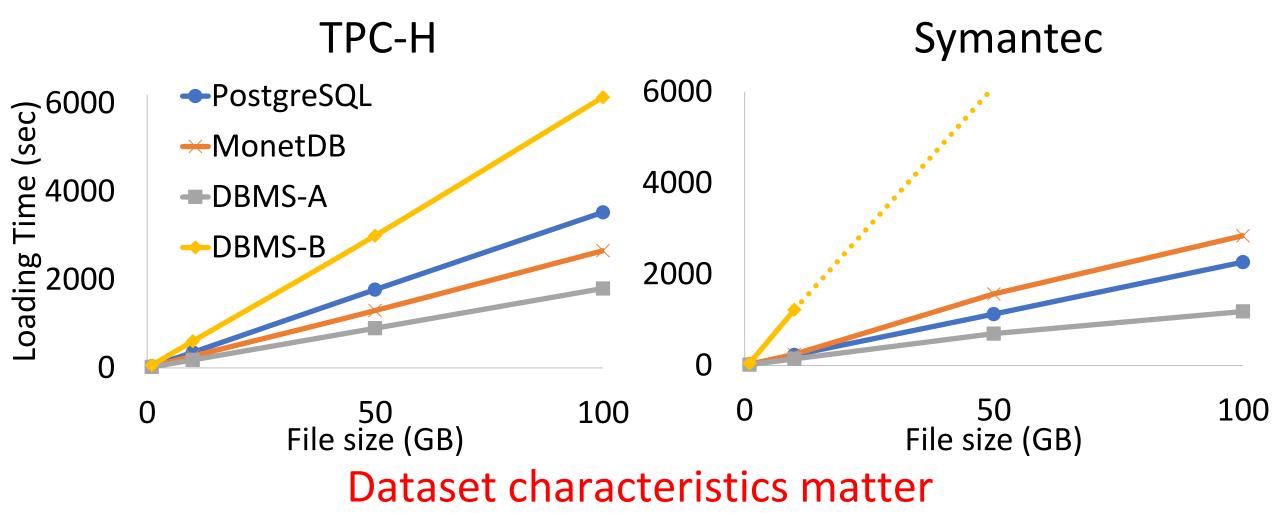
Data migration can be improved

Polystore: "One size does not fit all"

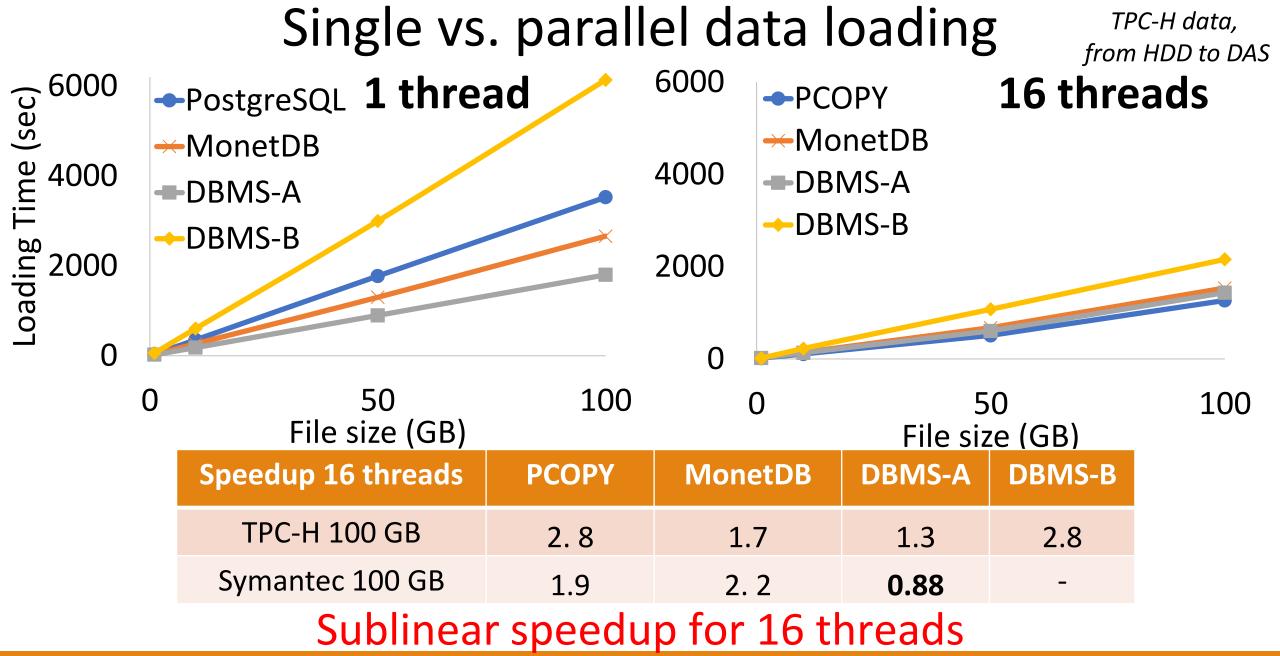


Polystore couples diverse data models

Single-threaded data loading



Effect of compression

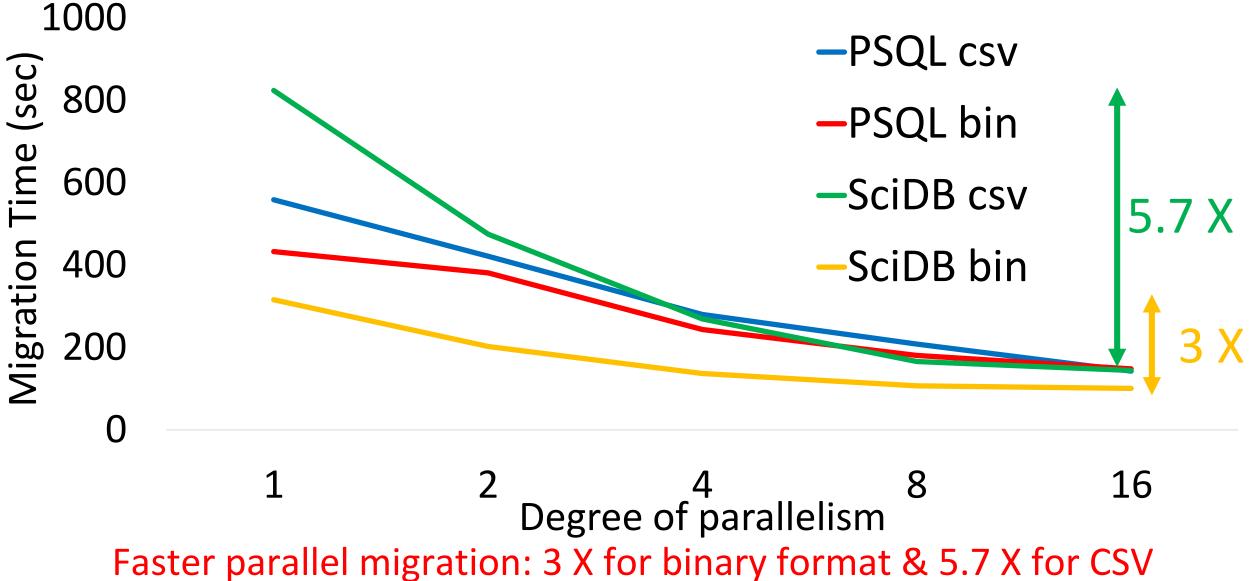


Data Migration from S-Store to PostgreSQL & SciDB

- Enhanced data export from S-Store
 - Binary PostgreSQL
 - Binary SciDB
- Parallel export via partitioning

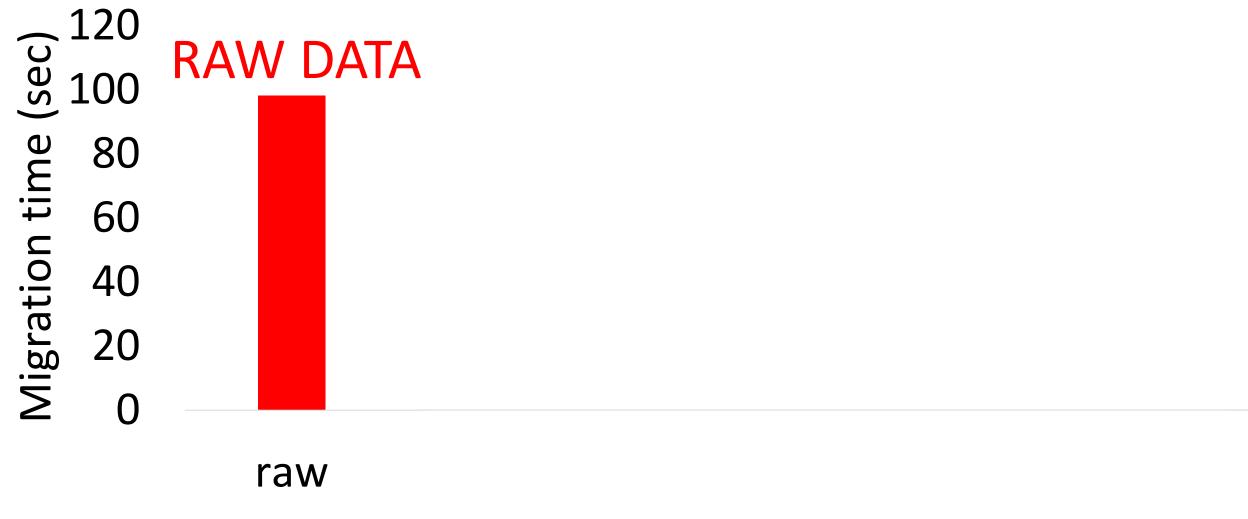


Data Migration from S-Store to PostgreSQL & SciDB



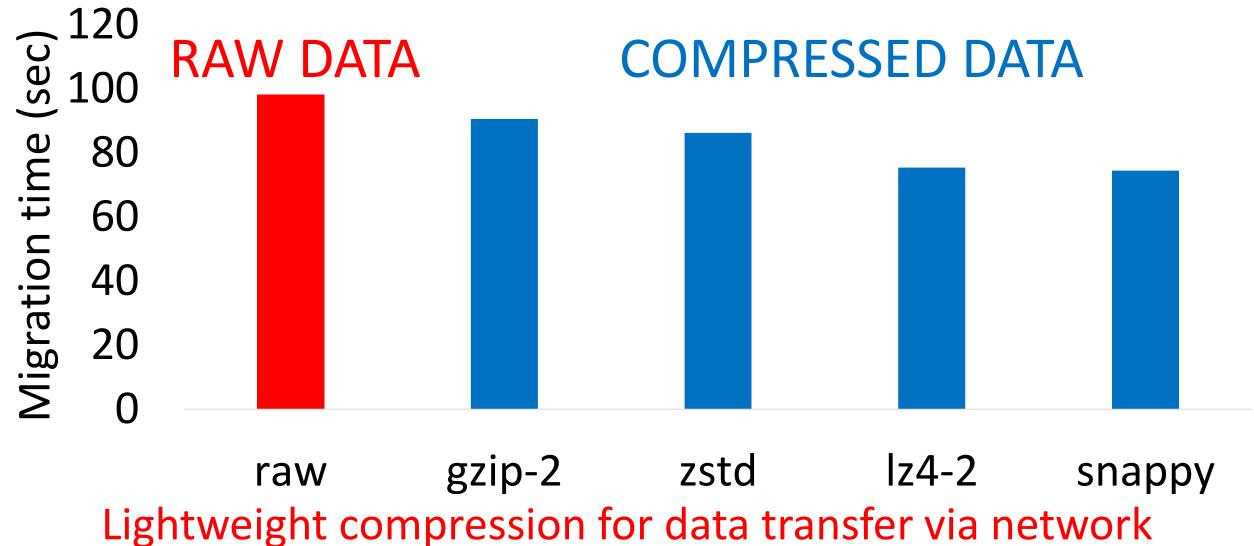
COMPRESSION for direct binary parallel migration

From PostgreSQL to SciDB, 4 threads, waveform data (int,int,double), 10 GB



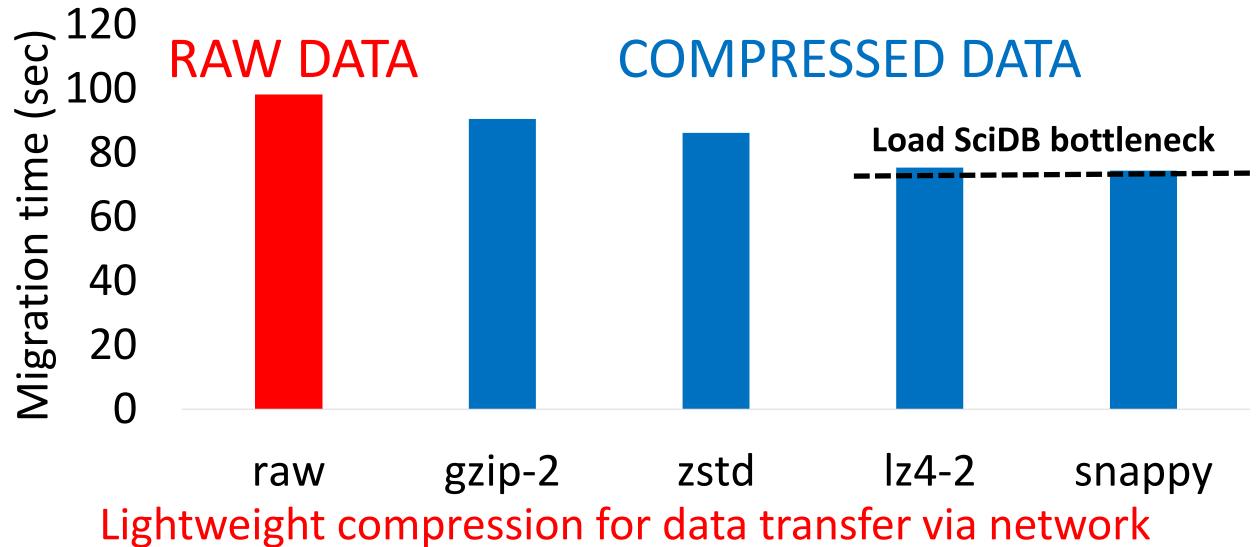
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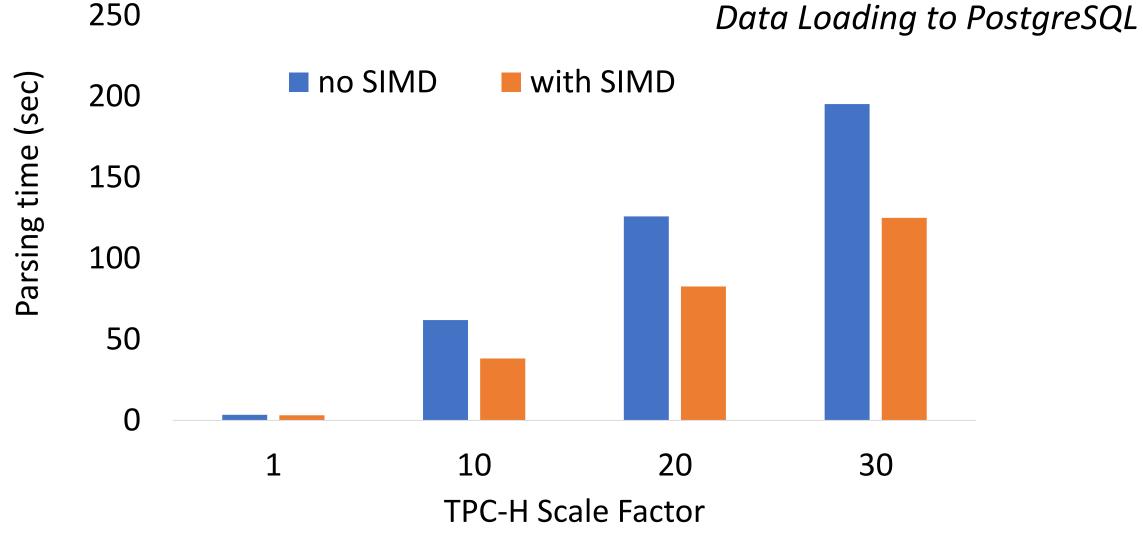


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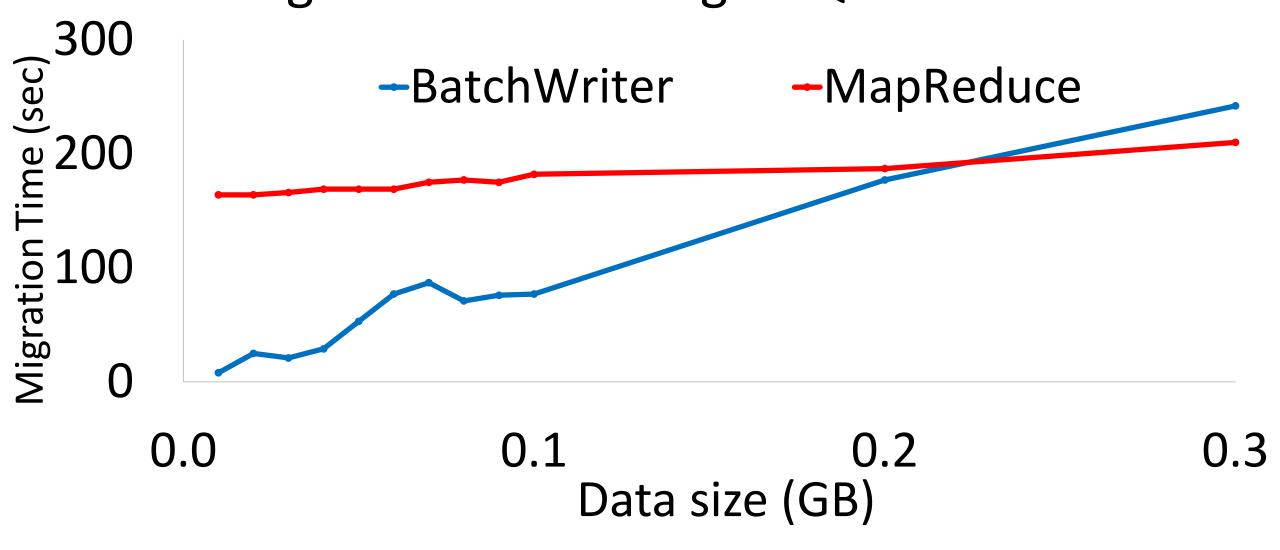


SIMD: for parsing the lines in input CSV file

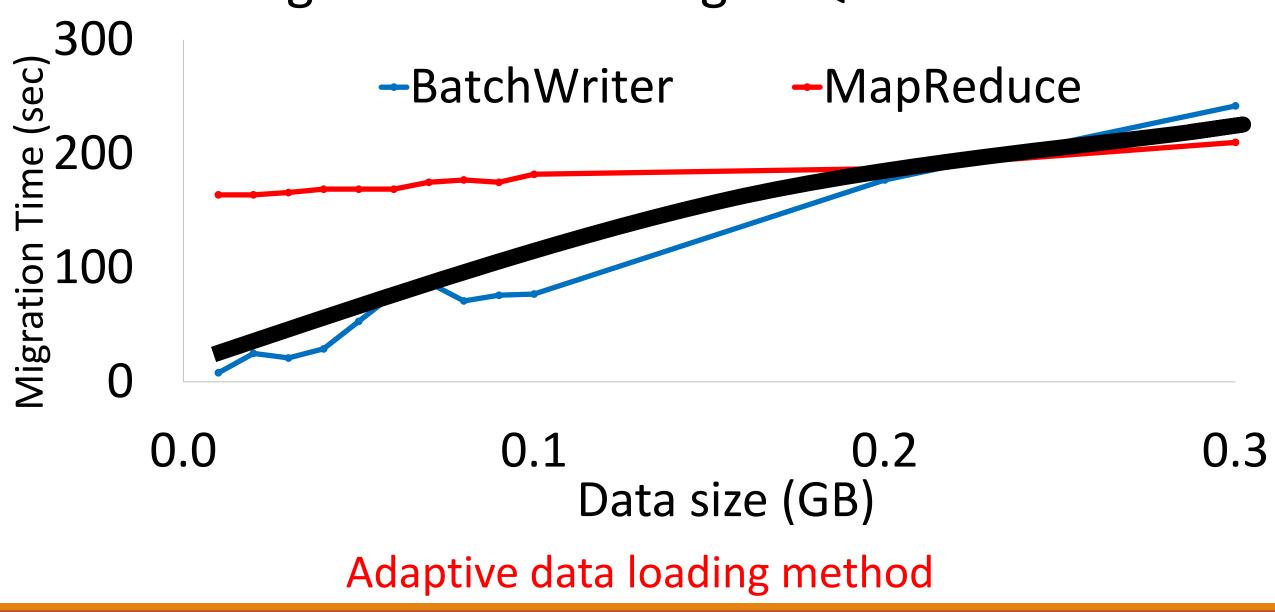


SIMD gives 1.6X speedup for parsing lines

Data Migration from PostgreSQL to Accumulo



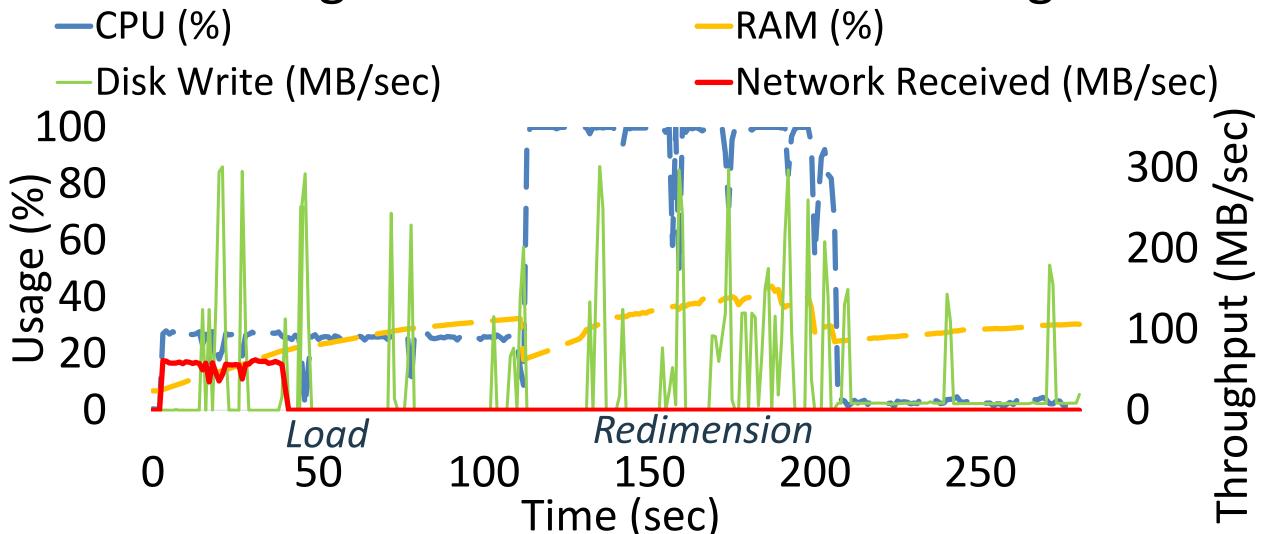
Data Migration from PostgreSQL to Accumulo

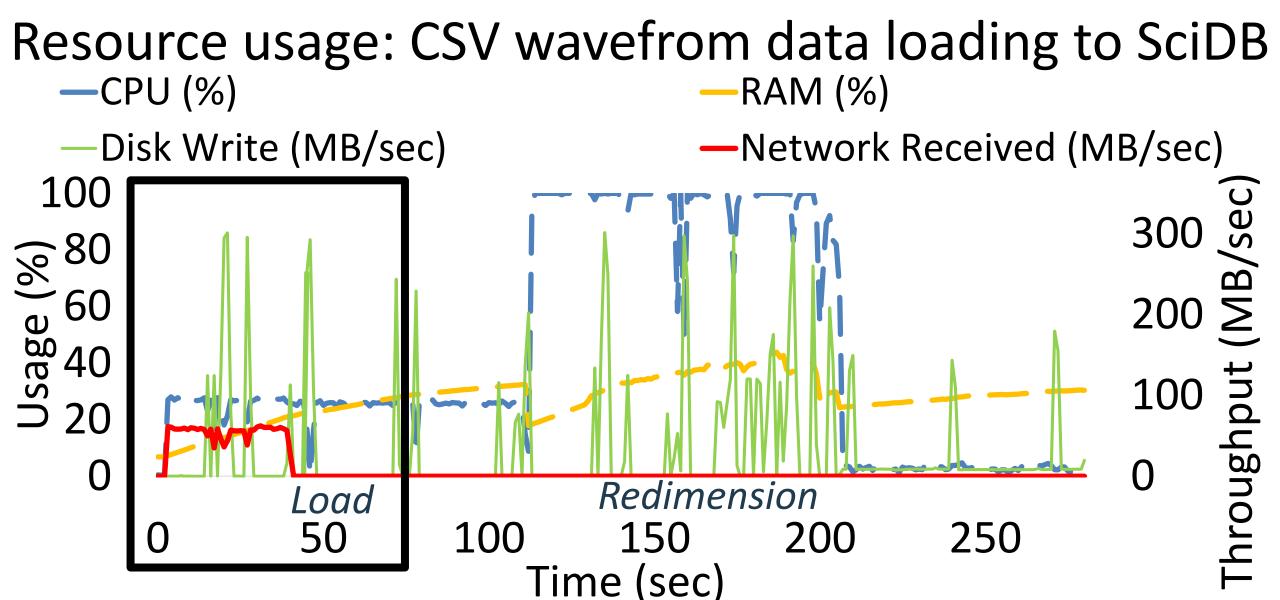


Future work

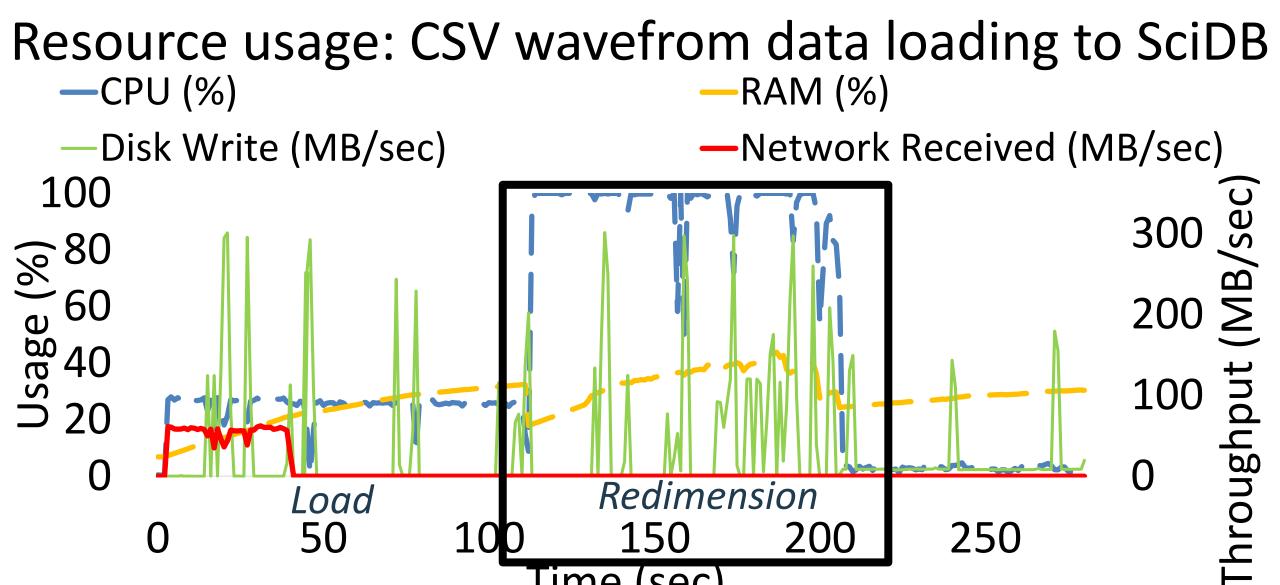
☐ Core migrator with a plug-in load/export interface for new DBMSs ☐ Common binary & SIMD-friendly format for DBMS data exchange ☐ Storage aware loading for the reading part & buffered writes handled in a separate thread (no bursts of writes) ☐ Equivalences between physical designs, e.g. mapping indexes in PostgreSQL to dimensions and chunk sizes in SciDB ☐ Improving the diverse analytics by integrating systems that are more homogenous for faster data sharing/migration with optimized code-paths for specialized analysis, e.g. SQL Server ☐ Integration with external Machine Learning tools (e.g. TensorFlow) and better data provisioning for the ML systems (e.g. via SQLite)

Resource usage: CSV wavefrom data loading to SciDB





Compress/Decompress to utilize spare CPU cycles



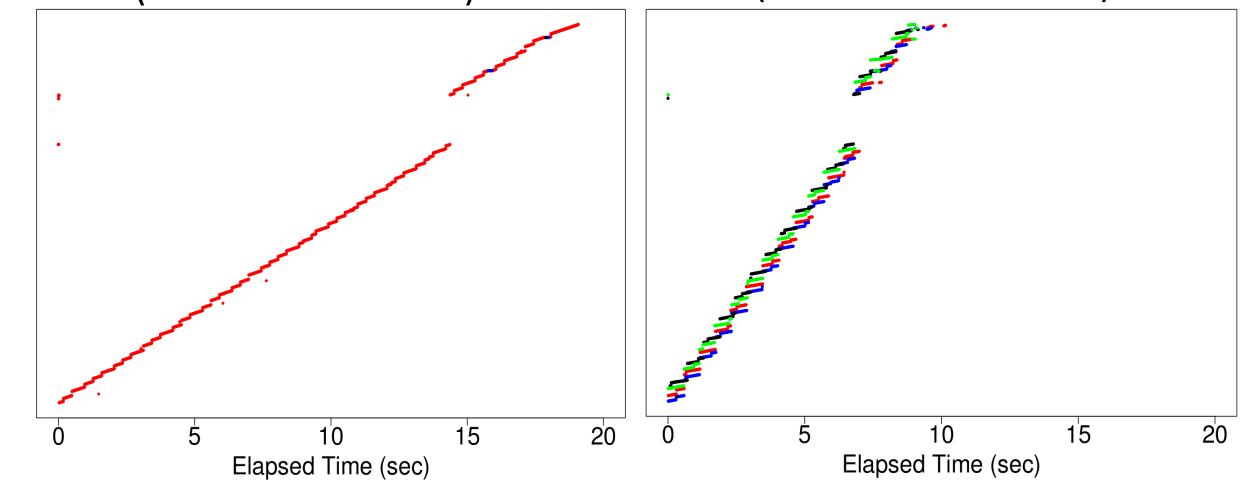
What is an optimal degree of parallelism?

Time (sec)

Single-threaded vs. Parallel Export from PostgreSQL



Block Address



Parallel loading in the presence of PK constraints

