# Parquet-MR acceleration using VectorAPI

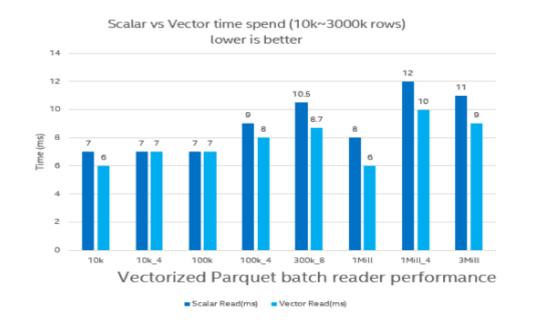
Jatin Bhateja (Intel) Fang Xie (Intel) Bitpack algorithm is very popular in database and big data storage system like Parquet, ORC. Vector API optimization provide better performance for en/decoding bitpack.

#### Below are performance impovement reports

Bit Width	1	2	3	4	5	6	7	8	9	10
Scalar time (ms):	3193	163	145	81	99	99	100	38	91	96
Vector time (ms):	873	28	41	15	25	19	17	8	12	12
Performance gain	3.66x	5.82x	3.53x	5.4x	3.96x	5.21x	5.88x	4.75x	7.58x	8.0x

#### Vectorized Execution: bit-packing decode

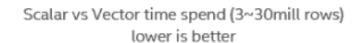
#### 

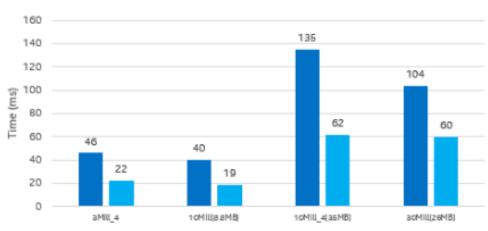


### Vectorized Execution: bit-packing decode

Test2: Reader test

- Location: Spark VectorizedParquetRecordReader
- Read mode: Batch column reader
- Bit width: 7
- Compression: false
- Batch Size: 4096
- ❖ Column: 1~4 columns
- Rows: 3Mills~30Mills

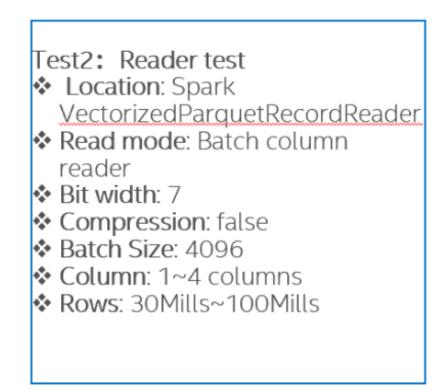


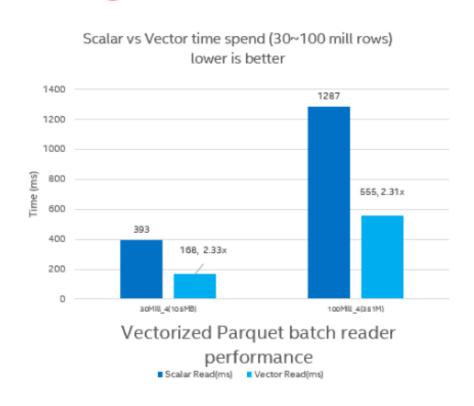


Vectorized Parquet batch reader performance

Scalar Read(ms) Vector Read(ms)

#### Vectorized Execution: bit-packing decode





Test sql query in Spark3.3.0 with OpenJDK17, the bitpack optimization improved sql query by  $2x\sim3x$  in some special case (deep scan file cases)

#### Hardware configuration Architecture: x86 64 CPU op-mode(s): 32-bit, 64-bit Byte Order: Little Endian CPU(s): 112 On-line CPU(s) list: 0-111 Thread(s) per core: 2 Core(s) per socket 28 Socket(s): NUMA node(s): Vendor ID: GenuineIntel BIOS Vendor ID: Intel(R) Corporation CPU family: 106 Model: Model name: Intel(R) Xeon(R) Platinum 8372C CPU @ 3.20GHz BIOS Model name: Intel(R) Xeon(R) Platinum 8372C CPU @ 3.20GHz Stepping: 3500.000 CPU MHz: CPU max MHz: 3500.0000 CPU min MHz: 800.0000 6400.00 BogoMIPS: Virtualization: VT-x L1d cache: 48K L1i cache: 32K L2 cache: 1280K L3 cache: 43008K NUMA node0 CPU(s): 0-27.56-83 NUMA node1 CPU(s): 28-55.84-111

# Spark configuration Standalone cluster 1 host 1 master

1 worker Master memory: 2g Executor memory:32g Driver memory:5g

select count(\*) from <a href="mailto:ParquetTable">ParquetTable</a> where int32\_field>2000 and int32\_field1<2000

Test1 : SQL test

Location: Spark SQL

Data type: INT32

Compression: Snappy

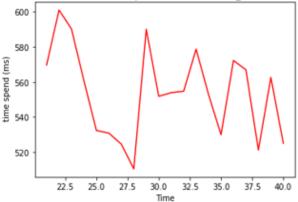
Column:4 columns

Rows: 100000000

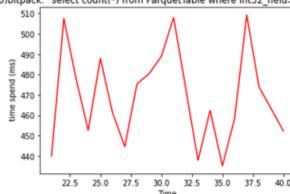
❖ File size: 652M

Perf gain: 15%

Scalar API v1 snappy (1~6400) for bitpack: "select count(\*) from ParquetTable where int32\_field>2000 and int32\_field1<2000 " everage time:554.0ms



Vector API optimization for v1 snappy (0~6400)bitpack: "select count(\*) from ParquetTable where int32\_field>2000 and int32\_field1<2000 " everage time:469.52ms



select sum(int32\_field2) from <a href="mailto:ParquetTable">ParquetTable</a> where int32\_field1 > 2000 group by int32\_field2

Test1 : SQL test

Location: Spark SQL

Data type: INT32

Compression: Snappy

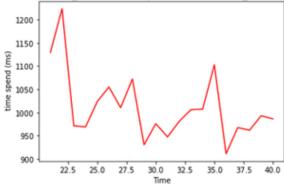
Column:4 columns

Rows: 100000000

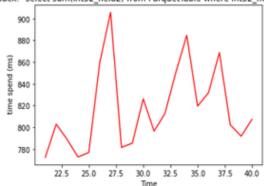
❖ File size: 652M

❖ Perf gain: 19%

Scalar API v1 snappy (1~6400) for bitpack: "select sum(int32\_field2) from ParquetTable where int32\_field1 > 2000 group by int32\_field2" average time:1011.31ms



Vector API optimization for v1 snappy (0~6400)bitpack: "select sum(int32\_field2) from ParquetTable where int32\_field1 > 2000 group by int32\_field2" everage time:816.92ms



select \* from ParquetTable where int32\_field1 =6399 order by int32\_field2

Test1: SQL test

Location: Spark SQL

Data type: INT32

Compression: Snappy

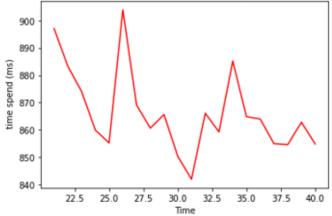
Column:4 columns

Rows: 100000000

❖ File size: 652M

Perf gain: 20%

Scalar API v1 snappy (1~6400) for bitpack: "select \* from ParquetTable where int32\_field1 =6399 order by int32\_field2" average time:866.38ms



Vector API optimization for v1 snappy (0~6400)bitpack: "select \* from ParquetTable where int32\_field1=6399 order by int32\_field2" everage time:675.79ms

