# Query Execution from Interpretation to Compilation on RDBMS

Driven by Hardware Revolution

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- Reviews
- Query Interpretation
- Query Compilation
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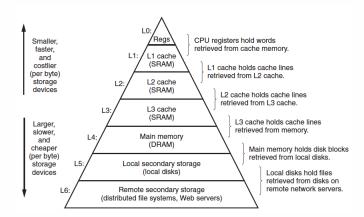


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# Storage hierarchy





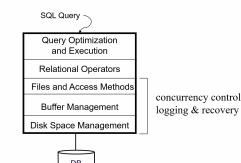
#### Architecture of Relational Database Management Systems



- Disk space management
- Buffer pool management
- Files and access methods
- Relational operators
- Query optimization and execution

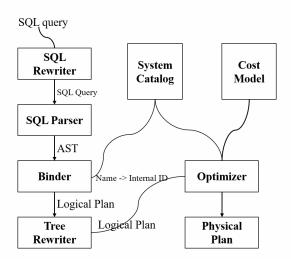
Lock Manager

System Catalog



#### Architecture of Query Planning





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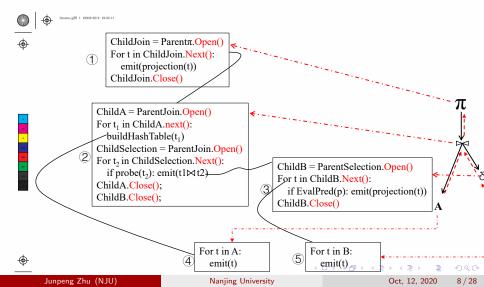
#### SQL Query Example



SELECT A.id , B. value FROM A , B WHERE A.id = B.id AND B. value >= 100 \* 3

# Iterator/Volcano Processing Model







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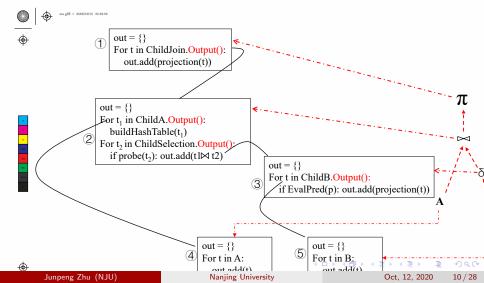
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# Materialization Processing Model







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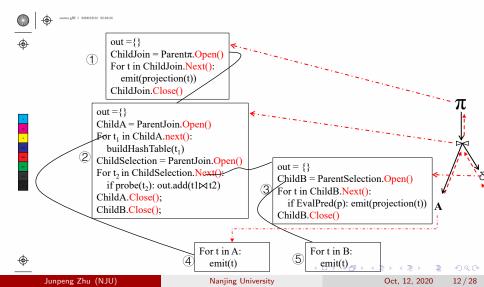
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# Vectorization Processing Model







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#### What's the problem?



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- CPU Inefficiency Iterator model of a large number of branch statements, memory access is not friendly to the CPU.

#### Attempt



Based on these issues, industry began to explore the path of query compilation, but it was not until recent years that it began to get more practical application.

Industrial products such as Impala, SparkSQL, PostgreSQL also began to adopt the scheme of compilation execution.

However, there is still a lot of imagination space for query compilation. And there is still a gap between the product landing in the industry and the academic research.

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#### Questions



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- What can be compiled in a query?
- How to compile?

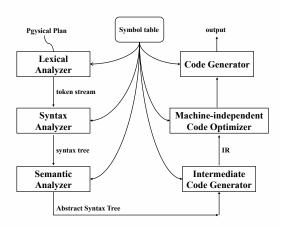
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- What can be compiled in a query?
- How to compile?
- What is the result of compiling?

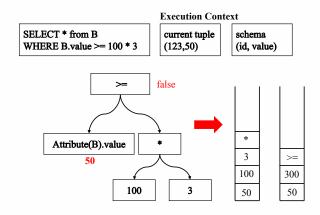
#### Compiler Overview





## Examples of Expression Interpretation





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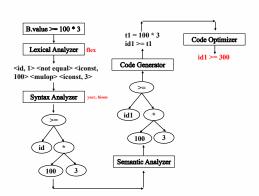
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How to solve these issues?



# Examples of Expression Compilation





## Example of Attributes of Tuples Compilation



```
void MaterializeTuple(char* tuple) {
  for (int i = 0; i < num_slots_; ++i) {
    char* slot = tuple + offsets_[i];
    switch(types_[i]) {
      case BOOLEAN:
        *slot = ParseBoolean();
        break;
      case INT:
        *slot = ParseInt();
        break;
      case STRING: ...
        // etc.
    }
}</pre>
```

interpreted

codegen'd

## **Example of Operators Compilation**



#### Interpretation

for t in range(table.num\_tuples):
 tuple = get\_tuple(table, t)
 if eval(predicate, tuple, parameters):
 emit(tuple)

#### Compilation

tuple\_size = xxx predicate.offset = xxx parameters\_value = xxx for t in range(table.num\_tuples): tuple = table.data + t \* table.size val = (tuple+predicate\_offset) + 1 if (val == parameter\_value): emit(tuple)

- Get schema in the system catalog for table.
- Calculate the offset based on the tuple size.
- Return the pointer to the tuple.

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- Tahboub, R.Y. and Rompf, T., 2020, June. Architecting a Query Compiler for Spatial Workloads. In Proceedings of the 2020 ACM SIGMOD International Conference on Management of Data (pp. 2103-2118).
- Kersten, T., Leis, V., Kemper, A., Neumann, T., Pavlo, A. and Boncz, P., 2018. Everything you always wanted to know about compiled and vectorized queries but were afraid to ask. Proceedings of the VLDB Endowment, 11(13), pp.2209-2222.
- Tahboub, R.Y., Essertel, G.M. and Rompf, T., 2018, May. How to architect a query compiler, revisited. In Proceedings of the 2018 International Conference on Management of Data (pp. 307-322).

## References II



- Menon, P., Mowry, T.C. and Pavlo, A., 2017. Relaxed operator fusion for in-memory databases: Making compilation, vectorization, and prefetching work together at last. Proceedings of the VLDB Endowment, 11(1), pp.1-13.
- Shaikhha, A., Klonatos, Y., Parreaux, L., Brown, L., Dashti, M. and Koch, C., 2016, June. How to architect a guery compiler. In Proceedings of the 2016 International Conference on Management of Data (pp. 1907-1922).
- Lang, H., Mühlbauer, T., Funke, F., Boncz, P.A., Neumann, T. and Kemper, A., 2016, June. Data blocks: Hybrid OLTP and OLAP on compressed storage using both vectorization and compilation. In Proceedings of the 2016 International Conference on Management of Data (pp. 311-326).

### References III



- Nagel, F., Bierman, G. and Viglas, S.D., 2014. Code generation for efficient query processing in managed runtimes. Proceedings of the VLDB Endowment, 7(12), pp.1095-1106.
- Wanderman-Milne, S. and Li, N., 2014. Runtime Code Generation in Cloudera Impala. IEEE Data Eng. Bull., 37(1), pp.31-37.

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# Thank you! Welcome for any questions!



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