61A Lecture 34

Monday, December 2

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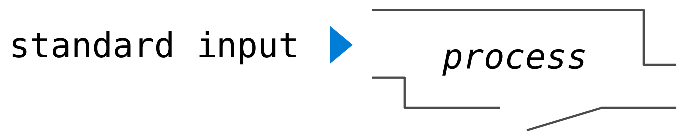

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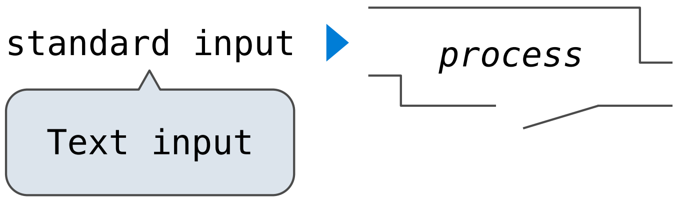

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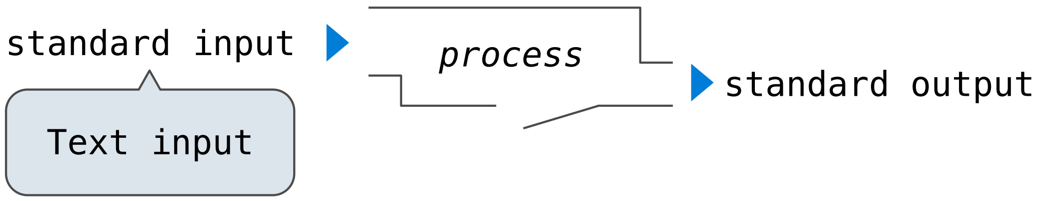

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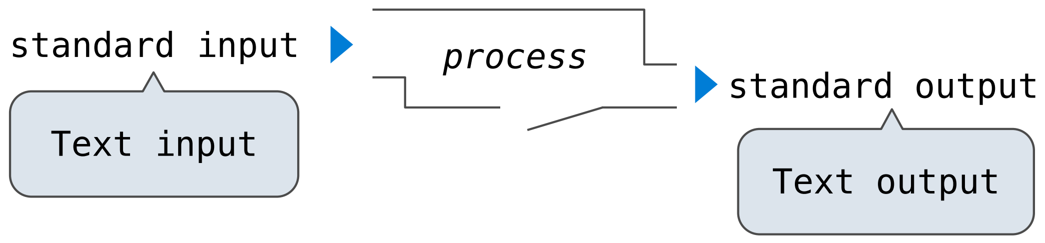

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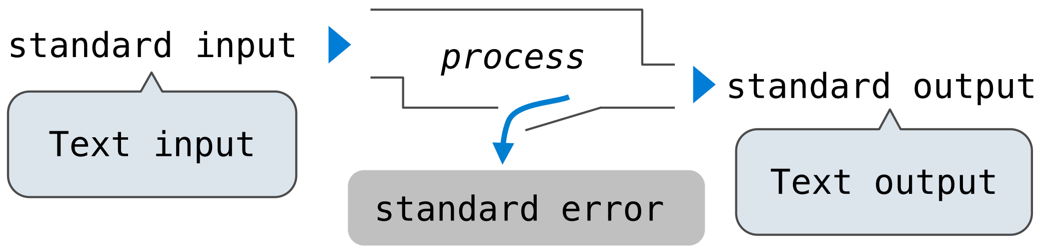

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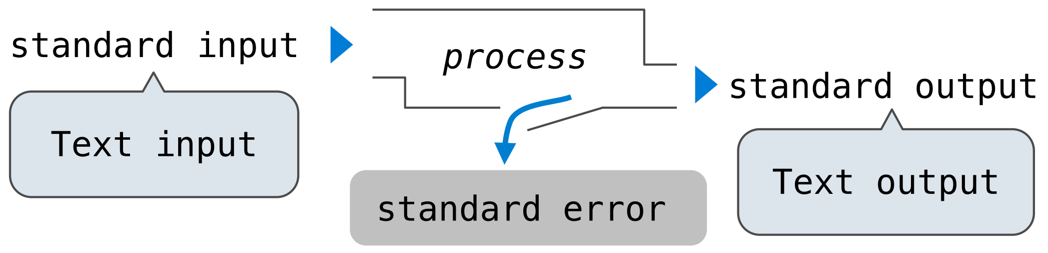

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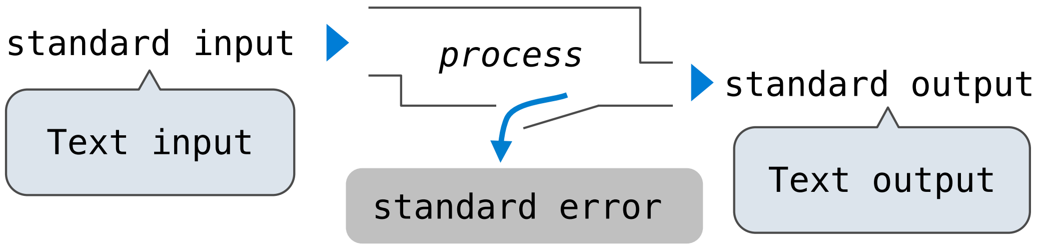

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(Demo)

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ls *.py | cut -f 1 -d '.' | grep hw | cut -c 3- | sort -n
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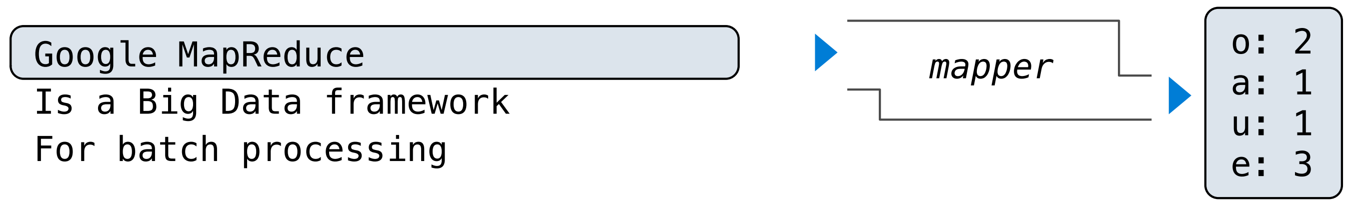

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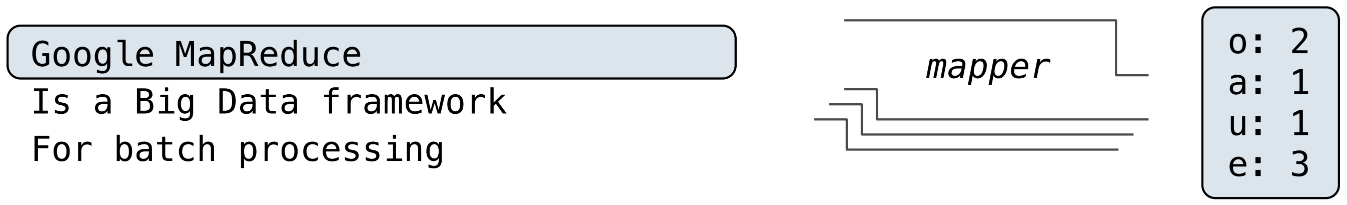

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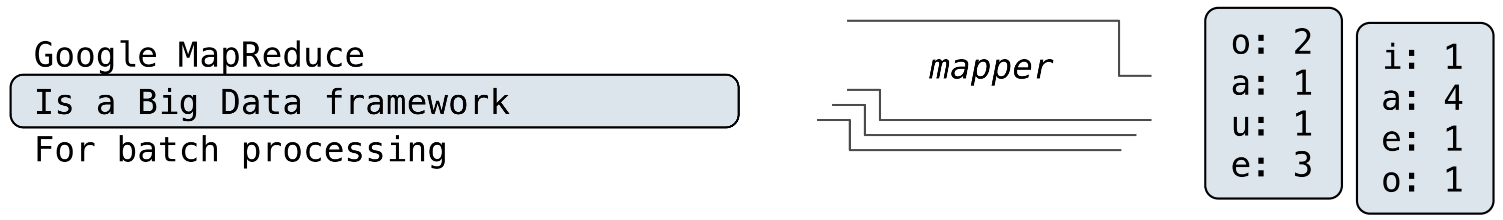

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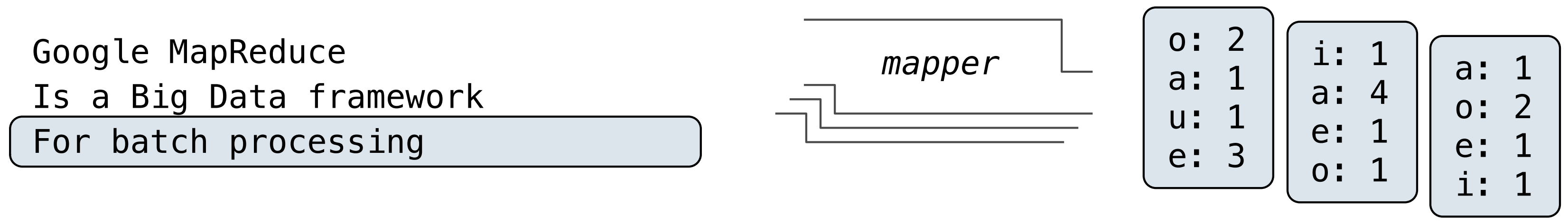

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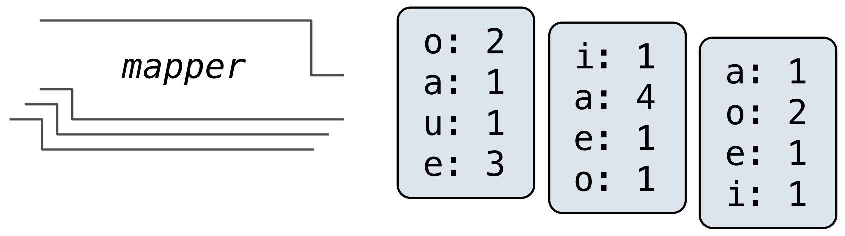

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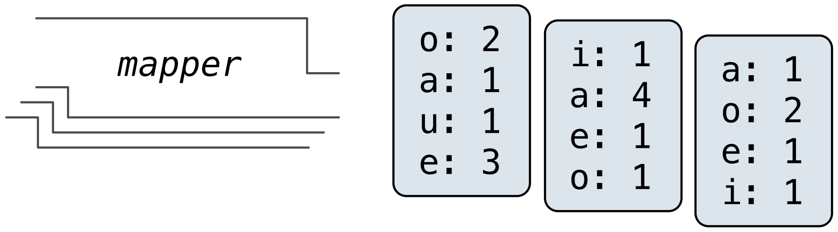

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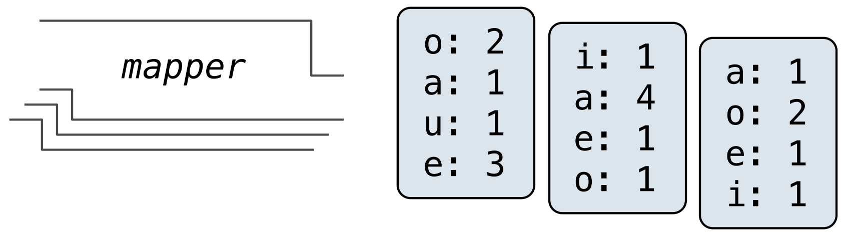

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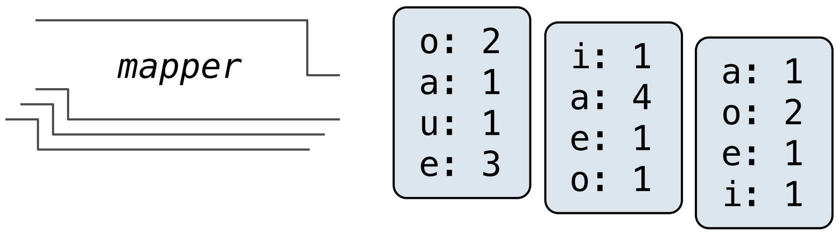

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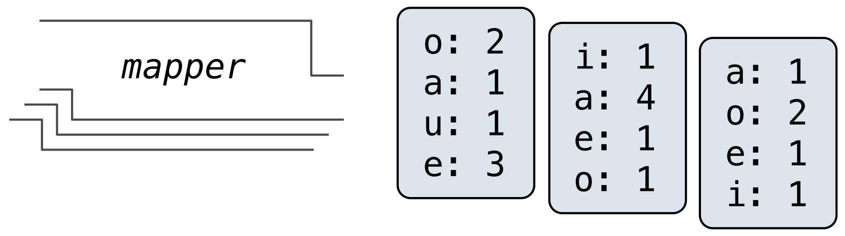

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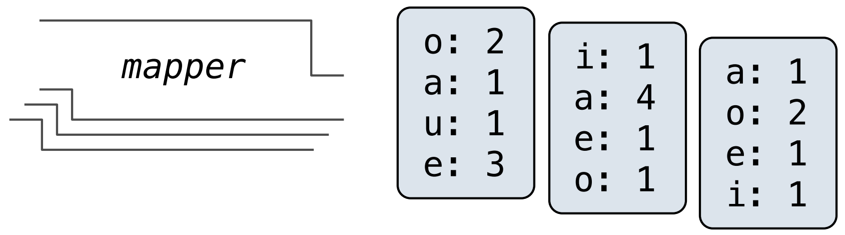

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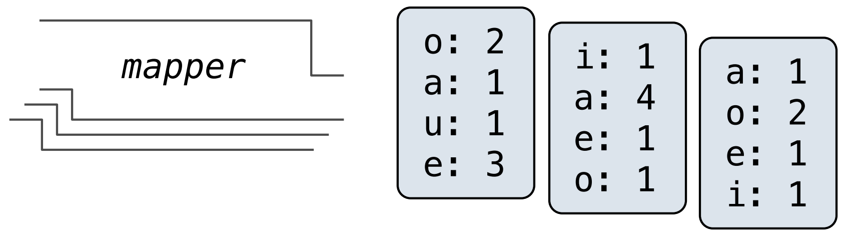

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```
a: 4
a: 1
a: 1
e: 1
e: 3
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...
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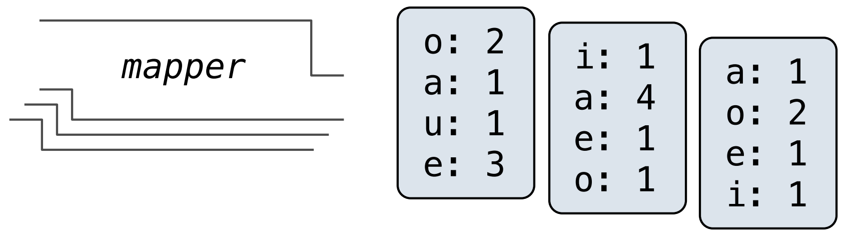

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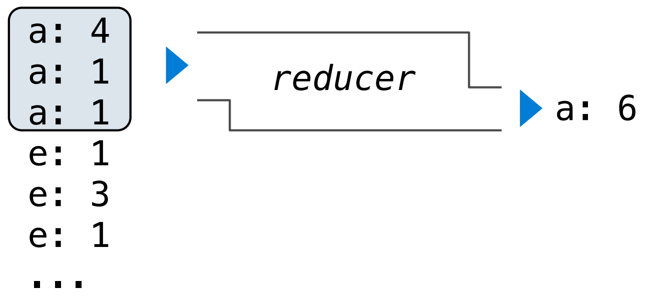

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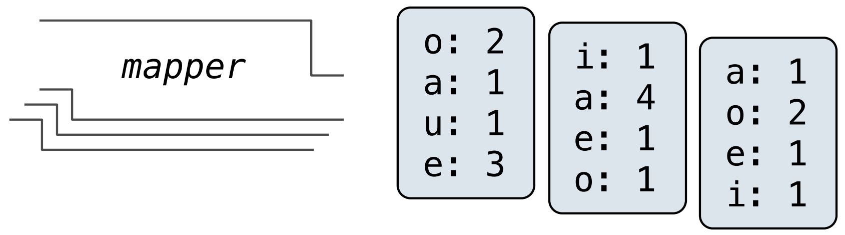

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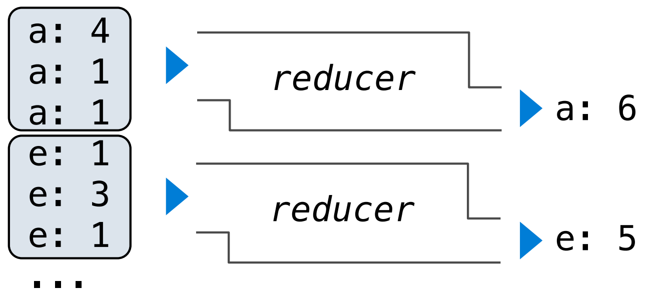

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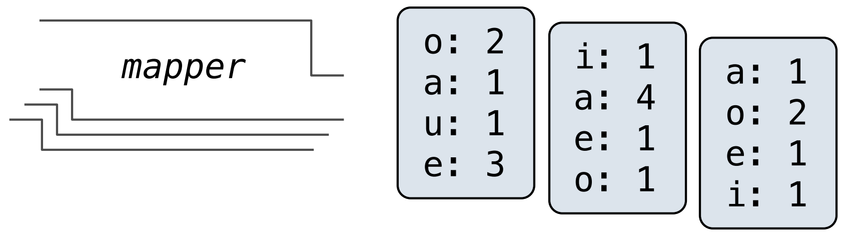

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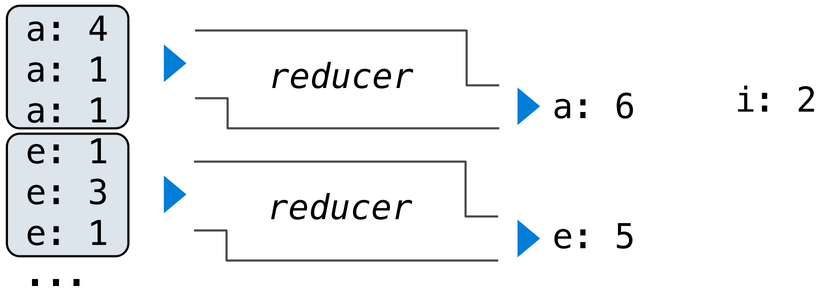

MapReduce Evaluation Model

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Google MapReduce
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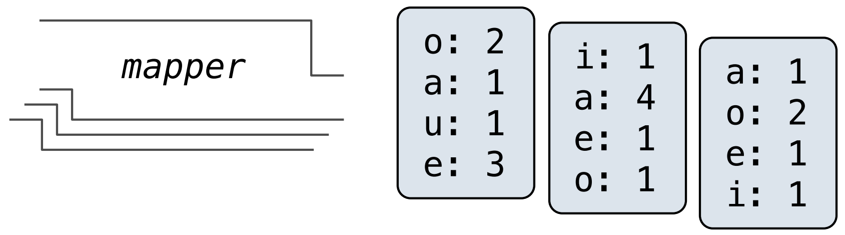

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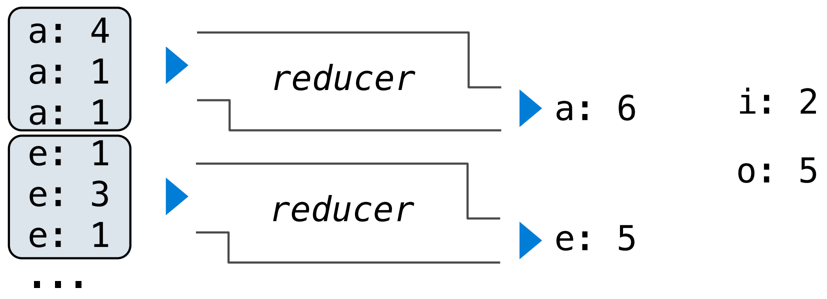

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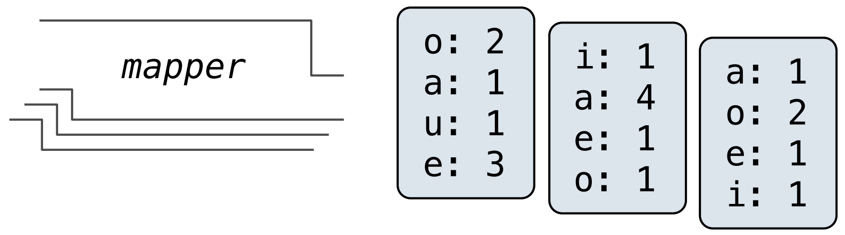

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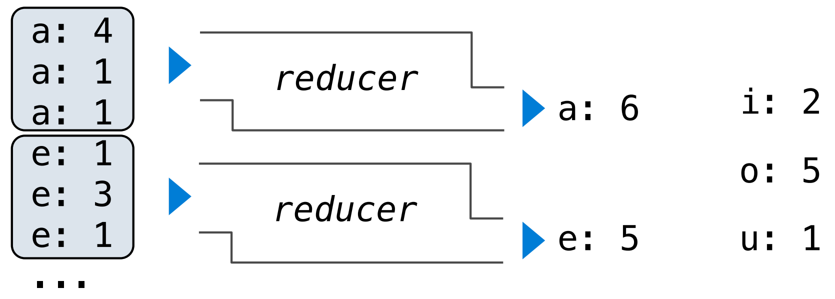

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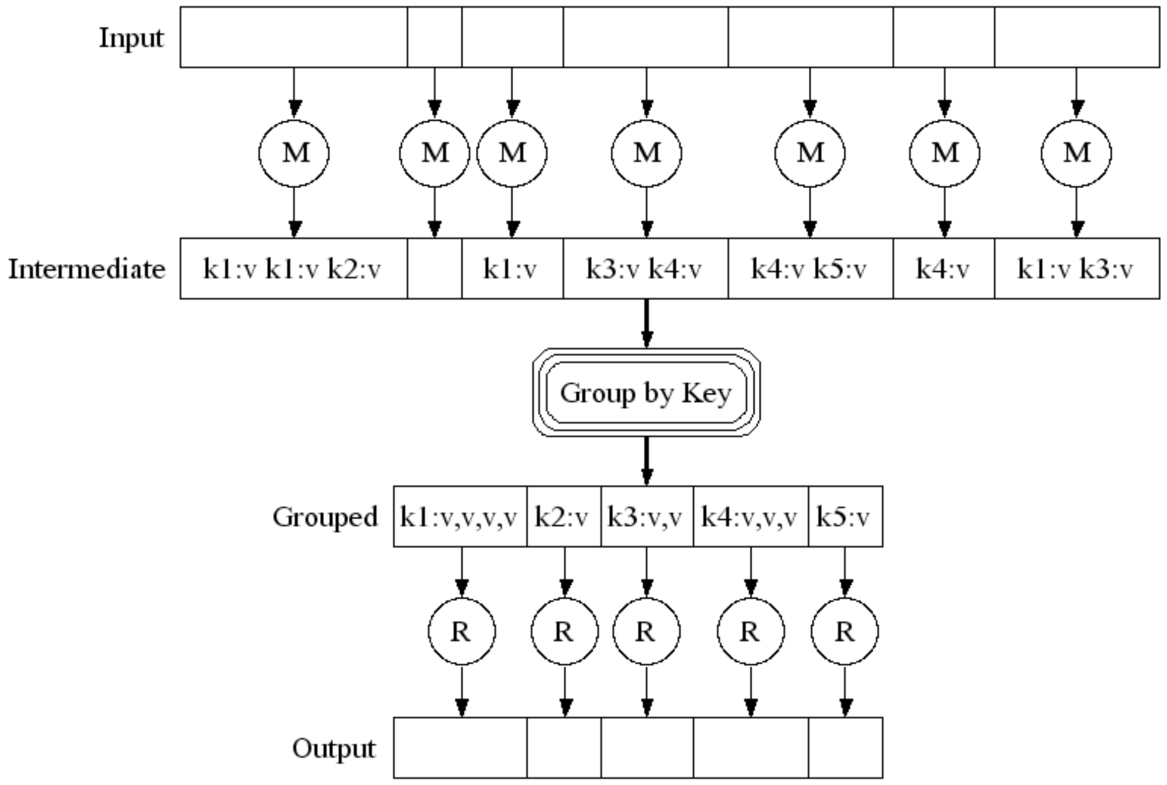
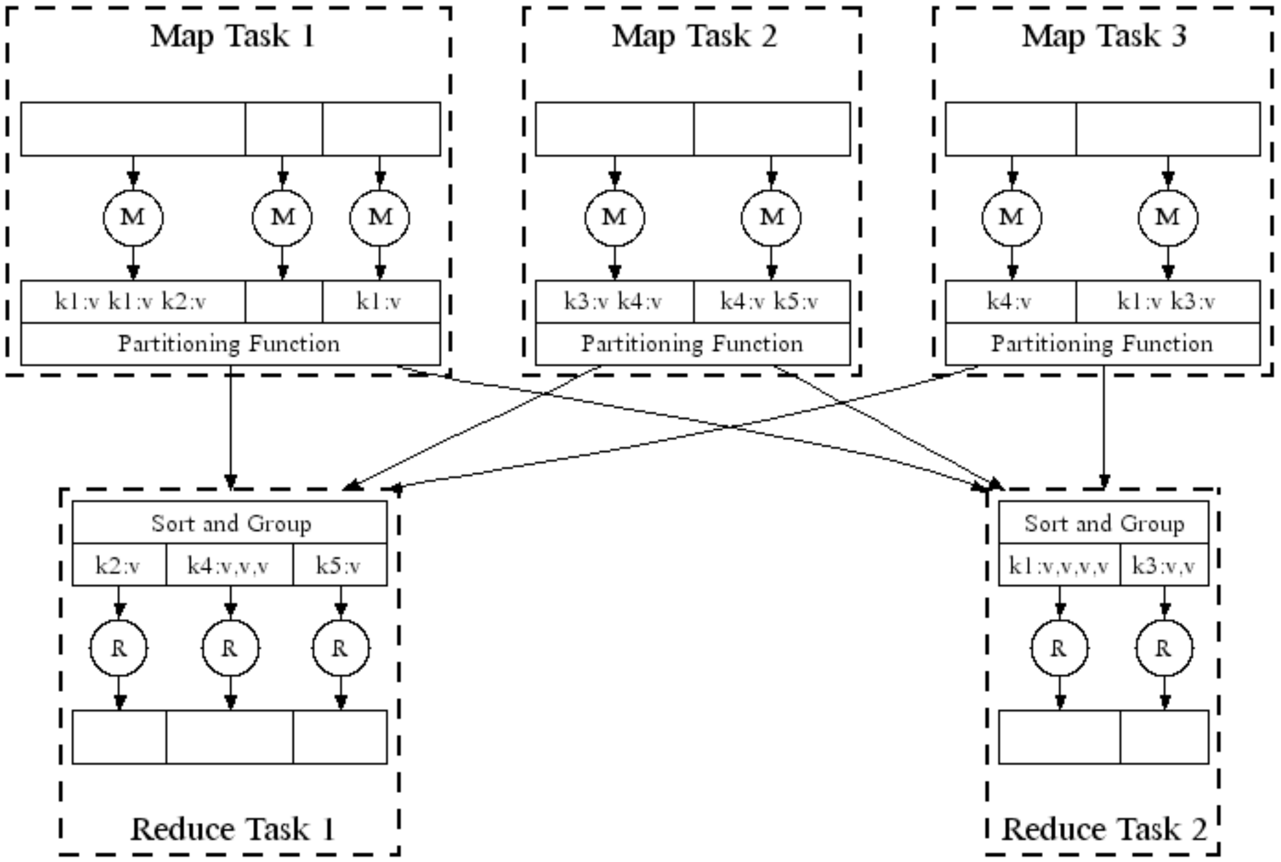
Execution Model



http://research.google.com/archive/mapreduce-osdi04-slides/index-auto-0007.html

Parallel Execution Implementation



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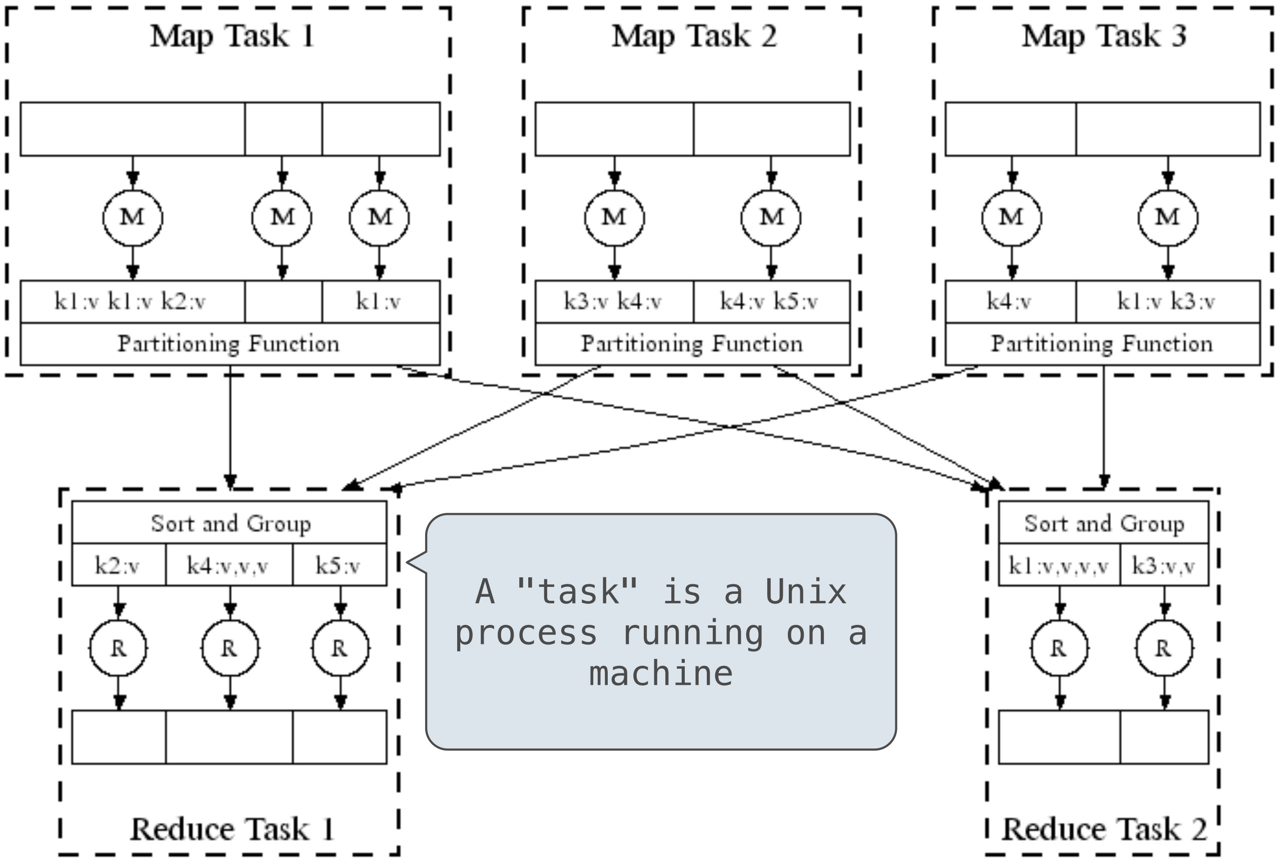

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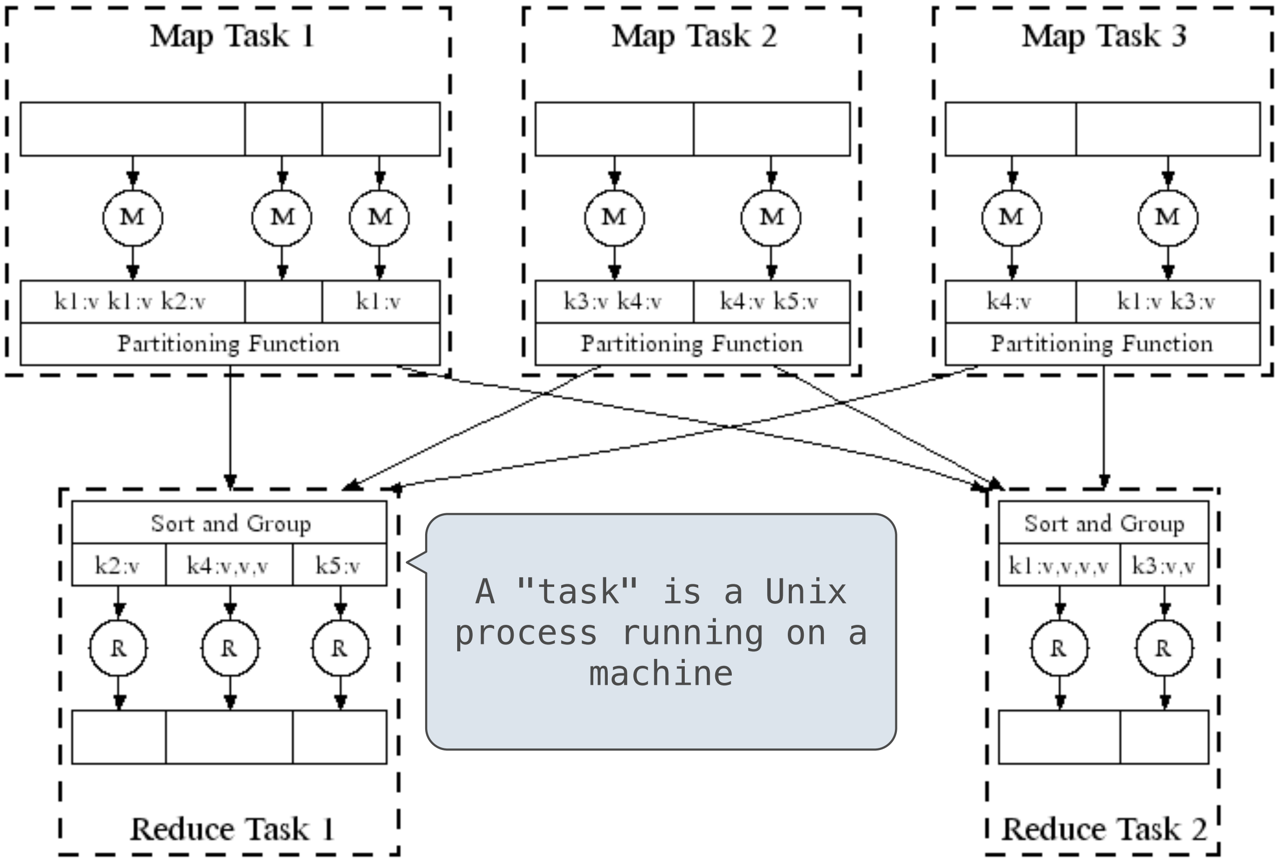

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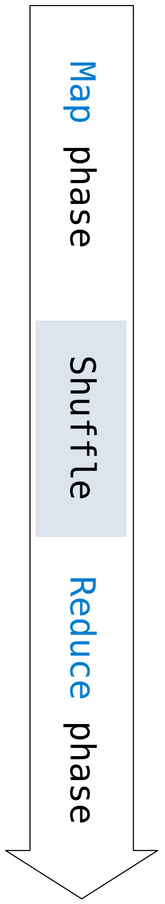

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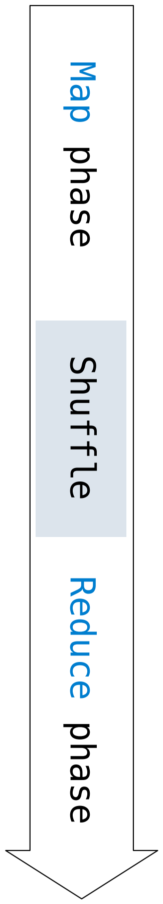

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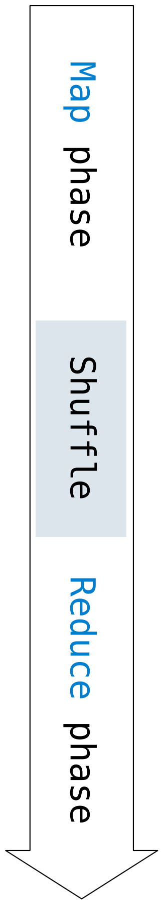

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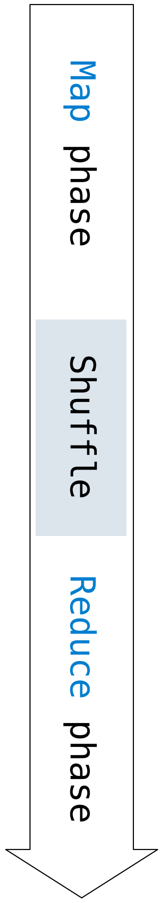

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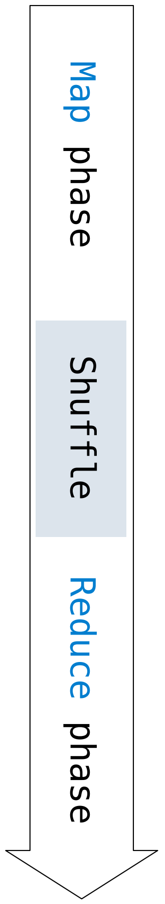

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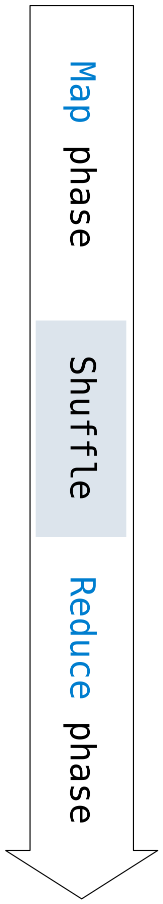

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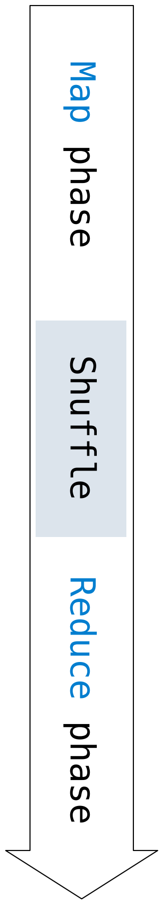

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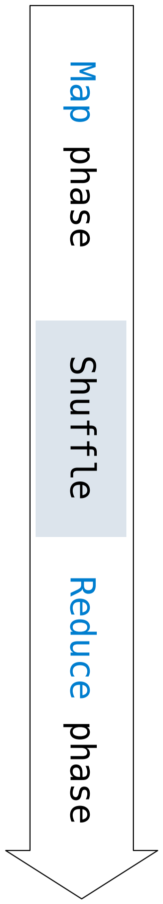

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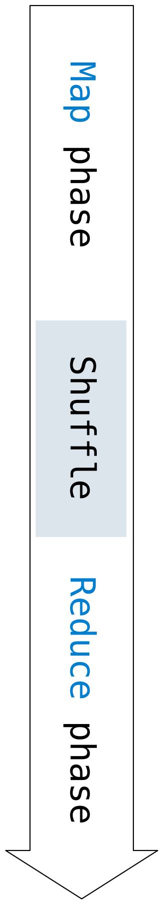

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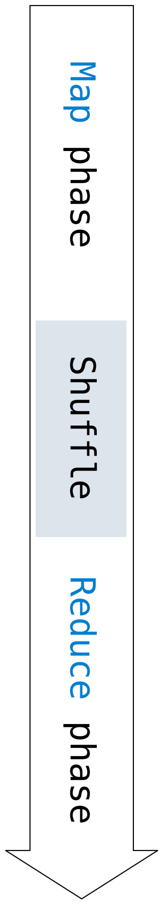

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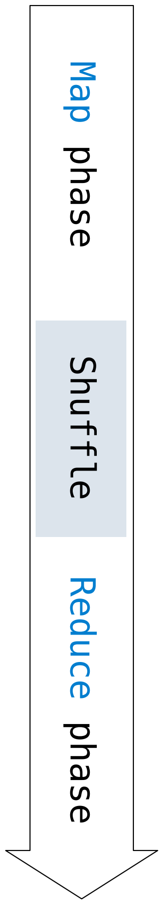

MapReduce Applications

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    emit(key, sum(value_iterator))
```

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(Demo)