

Theory of Map-Reduce Algorithms

Reducer Size

Replication Rate

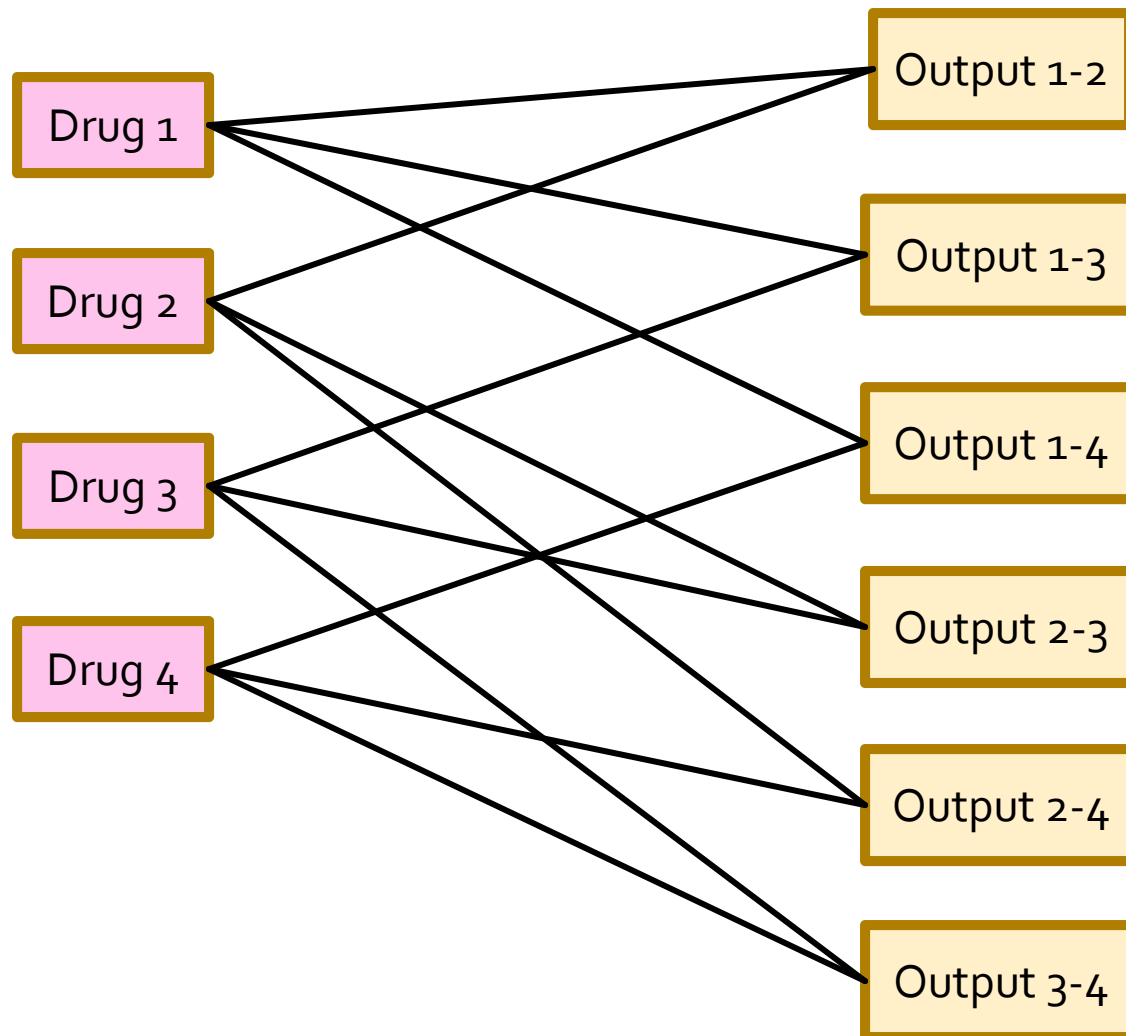
Mapping Schemas

Lower Bounds

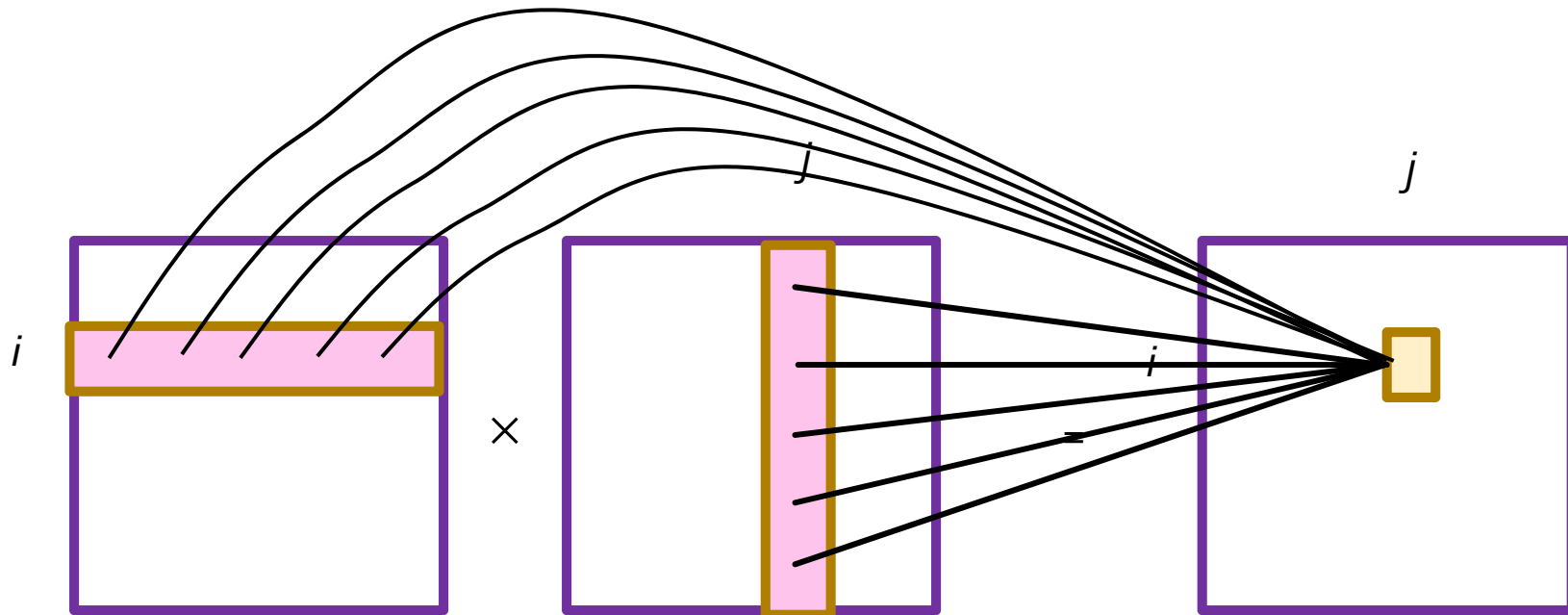
A Model for Map-Reduce Algorithms

1. A set of *inputs*.
 - **Example:** All drug numbers and their records.
2. A set of *outputs*.
 - **Example:** One output for each pair of drugs.
3. A many-many relationship between inputs and outputs.
 - An output is related to the inputs it needs to compute its values.
 - **Example:** The output for the pair of drugs $\{i, j\}$ is related to inputs i and j .

Example: Drug Inputs/Outputs



Example: Matrix Multiplication



Hypothetical and Actual Inputs

- In our model, inputs and outputs are hypothetical.
 - That is, they represent all the possible inputs that could exist, and the outputs that might be made.
- In any execution of the algorithm, the real inputs are a subset of the hypothetical inputs, and the outputs are whatever can be made from those inputs.

Example: Join

- Let's consider how we would model the join of two relations $R(A,B)$ and $S(B,C)$.
 - **Recall**: this is the set of tuples (a,b,c) such that (a,b) is in R and (b,c) is in S .
- Inputs = all $R(a,b)$ such that a and b are in the domains of attributes A and B , respectively + all $S(b,c)$ such that b and c are in the domains of attributes B and C .
- Outputs = all (a,b,c) such that a, b, c are in the domains of A, B, C .

Example – Continued

- The input/output relationship connects each output (a,b,c) with two inputs: $R(a,b)$ and $S(b,c)$.
- In any actual instance of the problem, only a small fraction of the hypothetical inputs will be present in the relations R and S .
- An output can be made if and only if both of its related inputs are present.

Reducer Size

- Two parameters tell a great deal about a map-reduce algorithm: *reducer size* and *replication rate*.
- Reducer size, denoted q , is the maximum number of inputs that a given reducer can have.
 - I.e., the length of the value list.
- Limit might be based on how many inputs can be handled in main memory.

Scaling Up Reducer Size

- When not all hypothetical inputs are expected to be present, we can raise q proportionally.
- **Example:** If we want no more than one million inputs to any reducer, and we expect 10% of inputs to be present, then we can set $q = 10M$.
- Risk of *skew*.
 - Input selection may not be random.
 - Some reducers may get many more inputs than others.

Replication Rate

- The average number of key-value pairs created by each mapper.
 - Denoted r .
- Represents the communication cost per input.
- If each reducer has size q , there are p reducers, and l is the number of inputs for the problem, then $r = pq/l$.

Example: Drug Interaction

- Suppose we use g groups and d drugs.
- A reducer needs two groups, so $q = 2d/g$.
- Each of the d inputs is sent to $g-1$ reducers, or approximately $r = g$.
- Eliminate g to get $r = 2d/q$.
- **Note:** since d is I , the number of inputs, and p , the number of reducers is approximately $g^2/2$, the relationship $r = pq/I$ holds as well.
 - That is, $r = (g^2/2)(2d/g)/d = g$.

Upper and Lower Bounds on r

- What we did gives an upper bound on the least possible r as a function of q .
 - A particular algorithm is surely no better than the best possible algorithm.
- A solid investigation of map-reduce algorithms for a problem includes lower bounds.
 - Proofs that you cannot have lower r for a given q .

Mapping Schemas

- A *mapping schema* for a problem and a reducer size q is an assignment of inputs to sets of reducers, with two conditions:
 1. No reducer is assigned more than q inputs.
 2. For every output, there is some reducer that receives all of the inputs associated with that output.
 - Say the reducer *covers* the output.

Mapping Schemas – (2)

- Every map-reduce algorithm has a mapping schema.
- The requirement that there be a mapping schema is what distinguishes map-reduce algorithms from general parallel algorithms.

Example: Drug Interactions

- d drugs, reducer size q .
- No reducer can cover more than $q^2/2$ outputs.
- There are $d^2/2$ outputs that must be covered.
- Therefore, we need at least d^2/q^2 reducers.
- Each reducer gets q inputs, so replication r is at least $q(d^2/q^2)/d = d/q$.
- Half the r from the algorithm we described.
 - We had $r = g$ and $q = 2d/g$ (or $g = r = 2d/q$).