



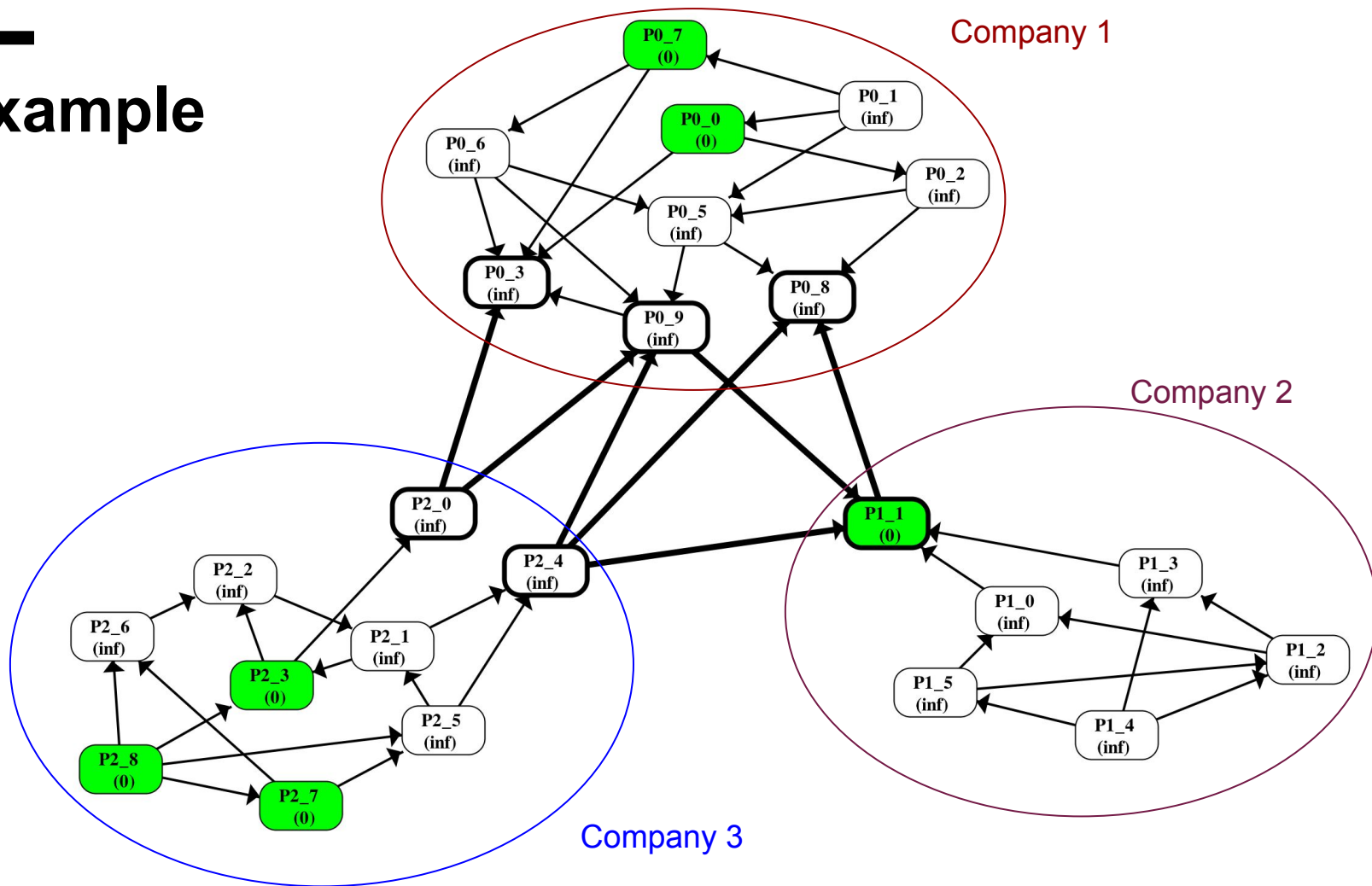
Network Distance

Fast Iterative MPC by Symbolic
Optimization of Unrolled Loops.

— Network Distance

- Large network with nodes belonging to different companies.
- Two kinds of nodes: Vulnerable, and Safe.
- Compute minimum distance to a vulnerable node for all nodes.
- Vulnerable nodes have distance 0
- Safe nodes have distance infinity.
- Computation consists of alternating rounds of addition and min.

Example



Security

- Passive Security.
- Hide internal topology of a company's network.
- Final distance is known but not any intermediate values.
- Final distance is known but not the causing vulnerable nodes.
- Gateways graph is public.
- Each company learns only the outputs of its own nodes.

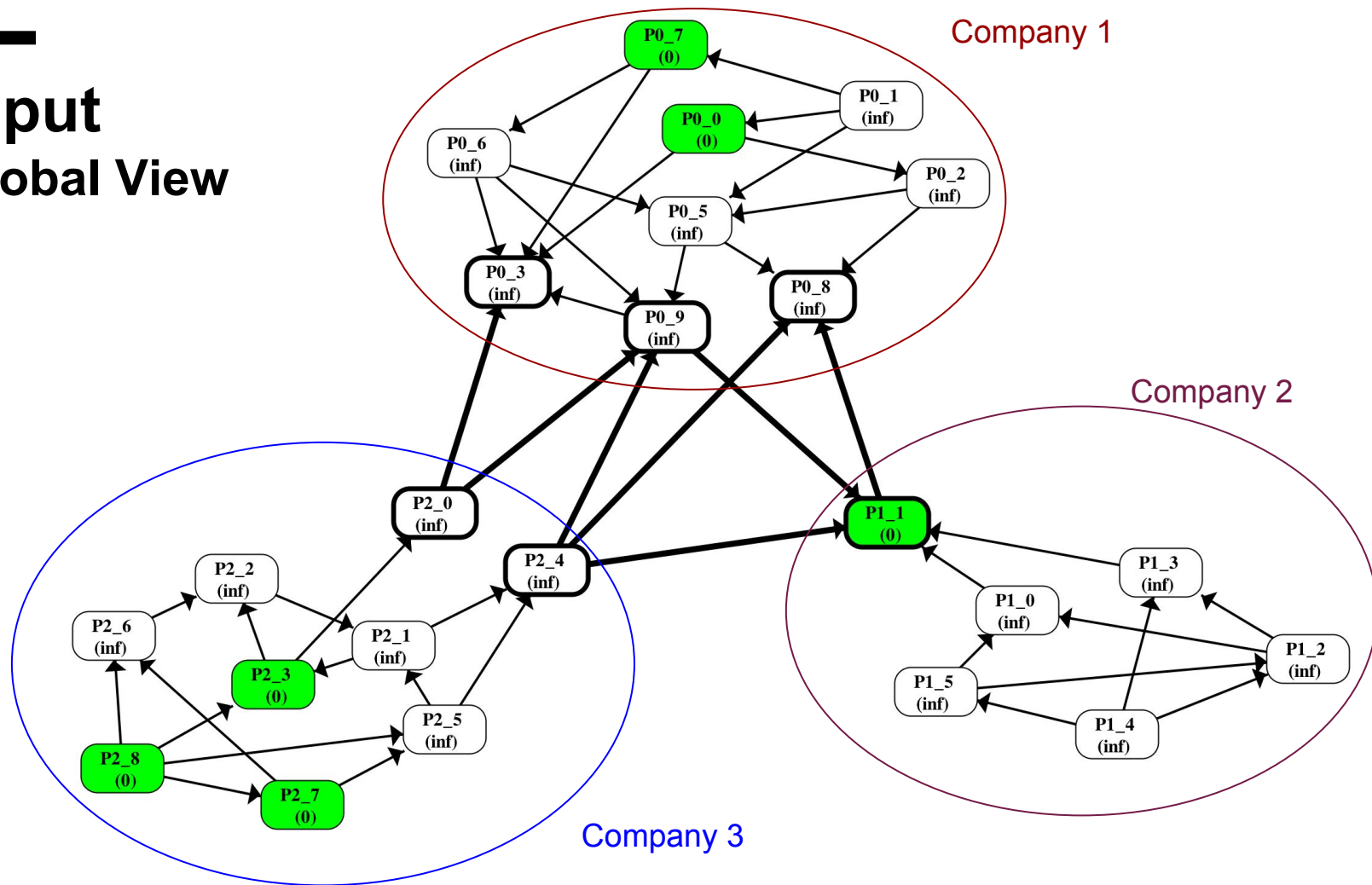
Problem

- Graph can be very large
 - Cannot perform MPC on the entire graph.

Solution

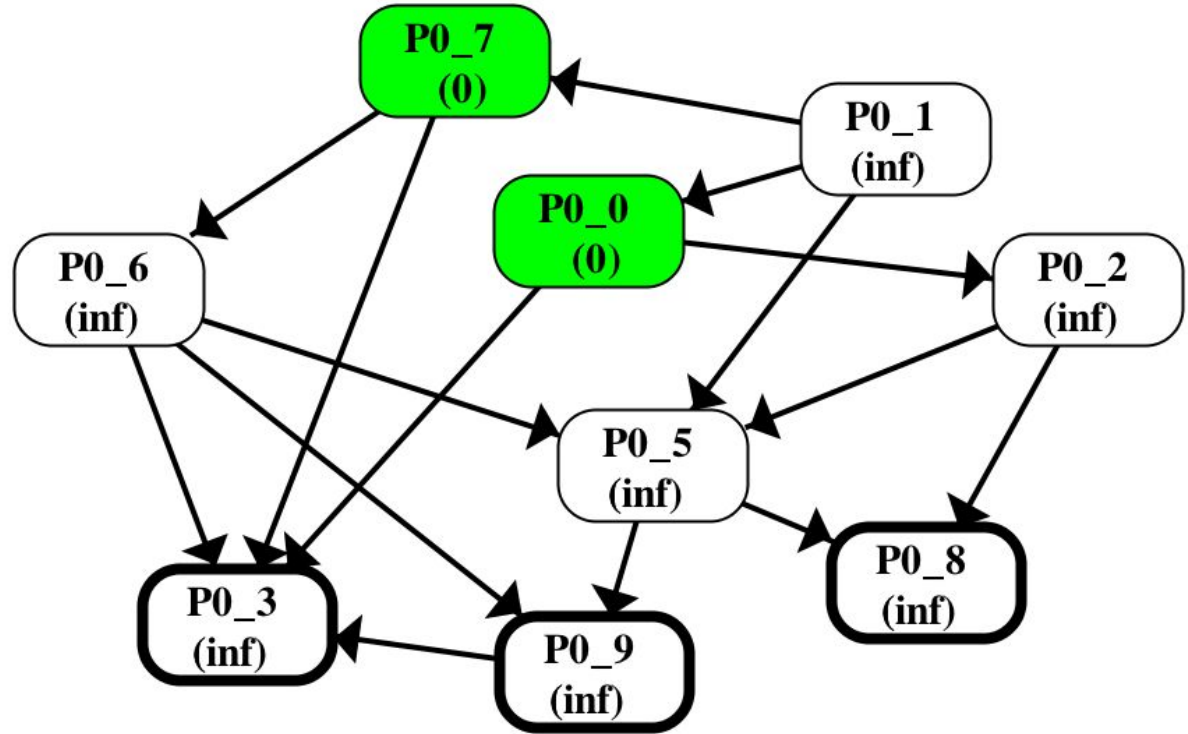
- Companies perform direct computation on their private network.
- MPC computation on the subgraph containing only gateways of companies.
- Direct computation to propagate gateways results back into private network.

Input Global View



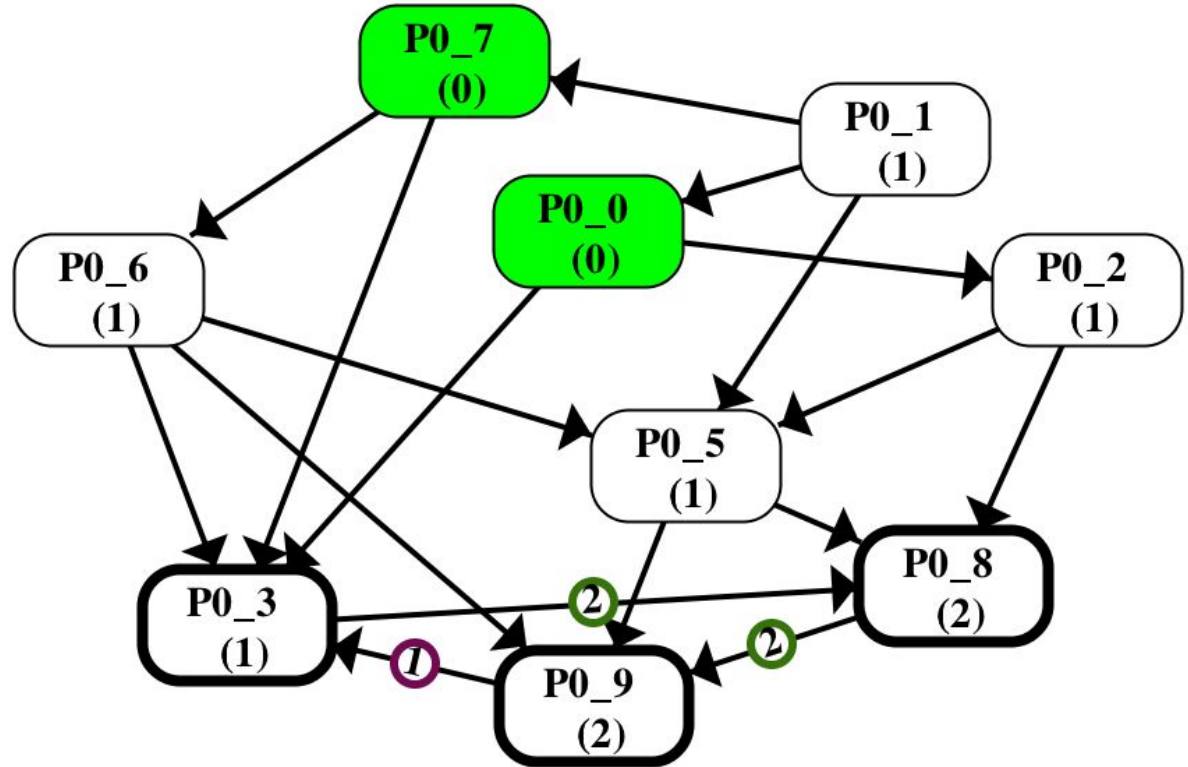
Local Network

Company 1



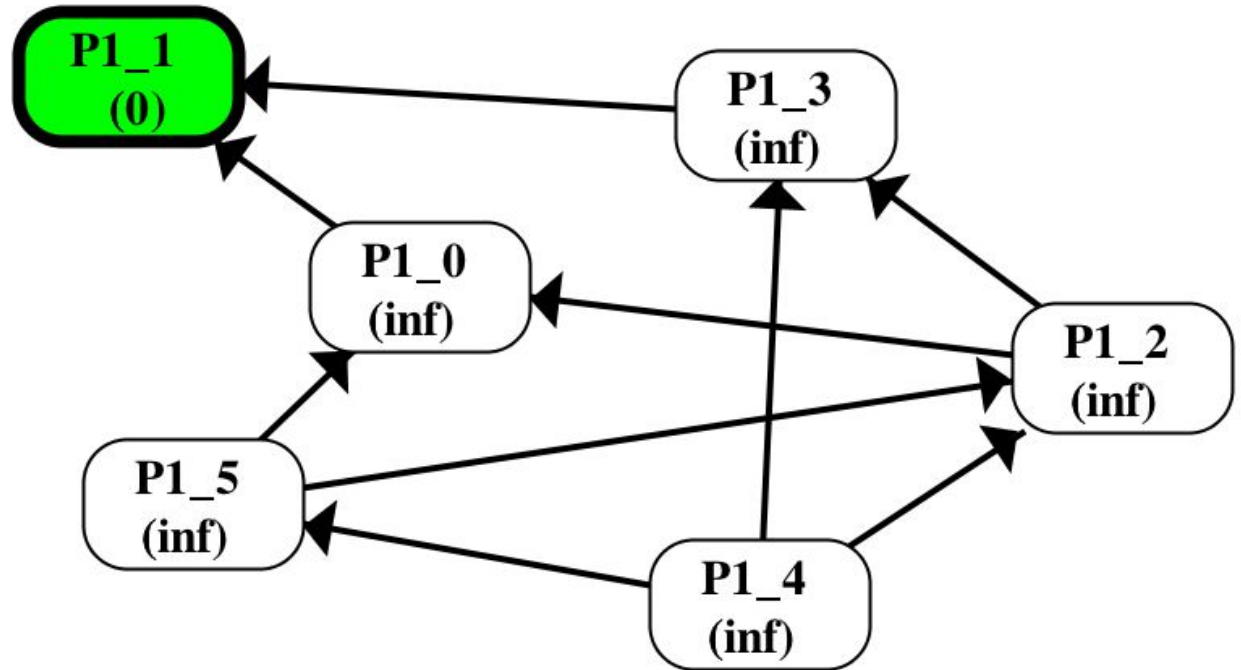
Local Computation Output

Company 1



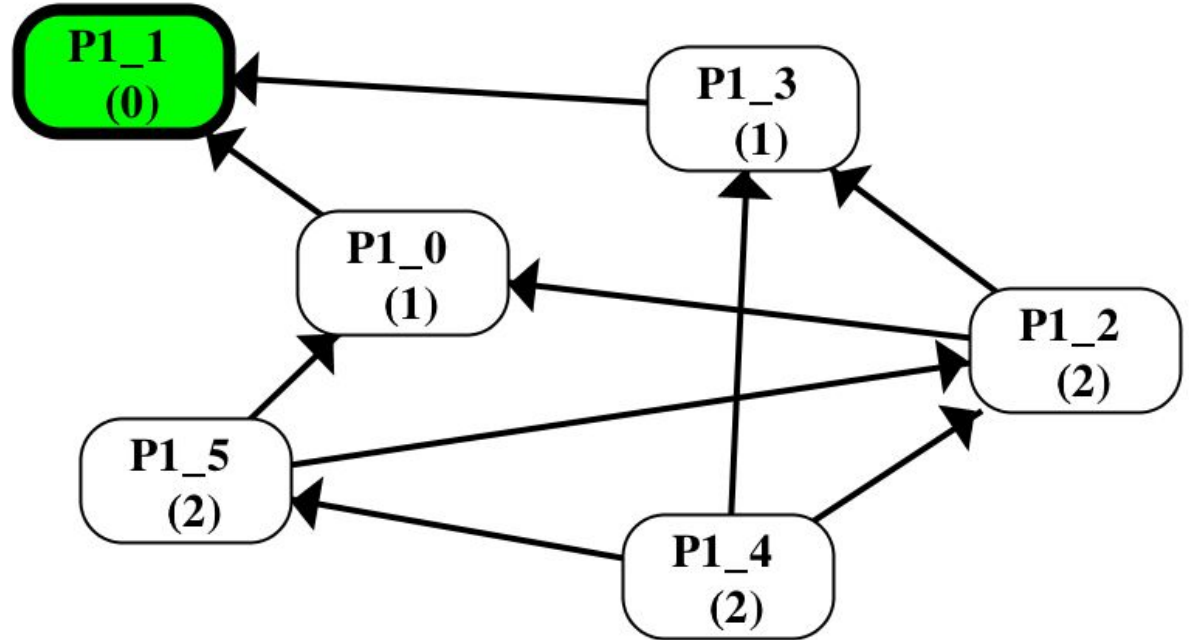
Local Network

Company 2



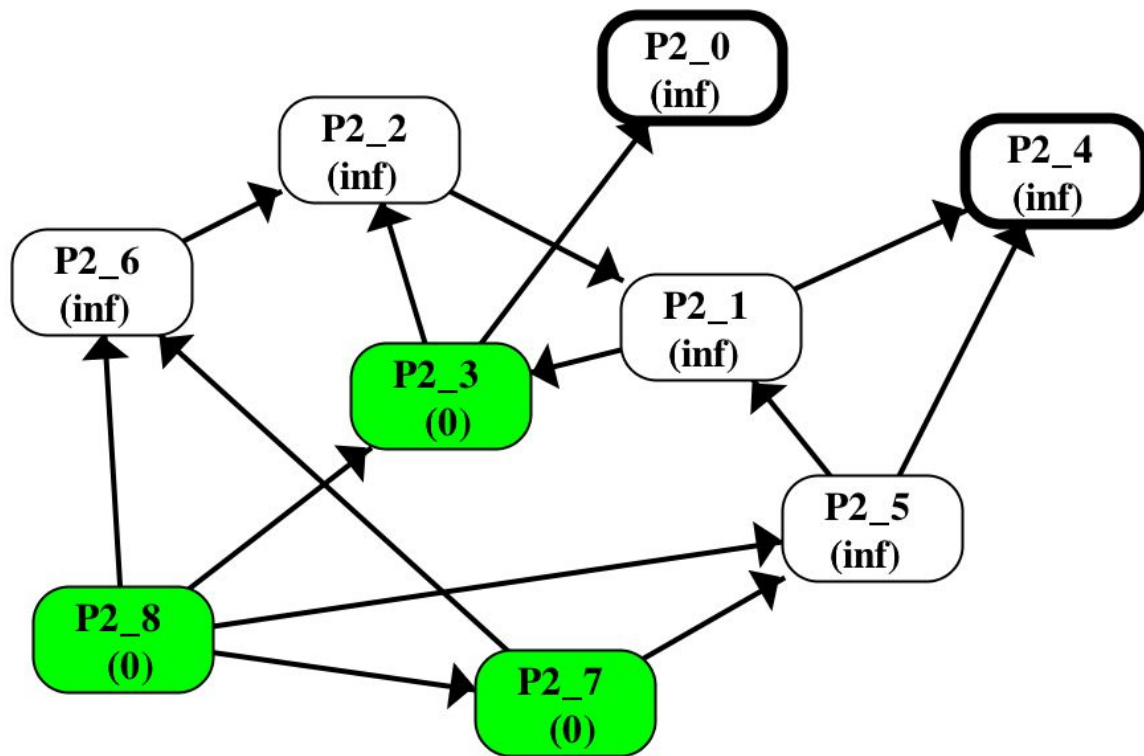
Local Computation Output

Company 2



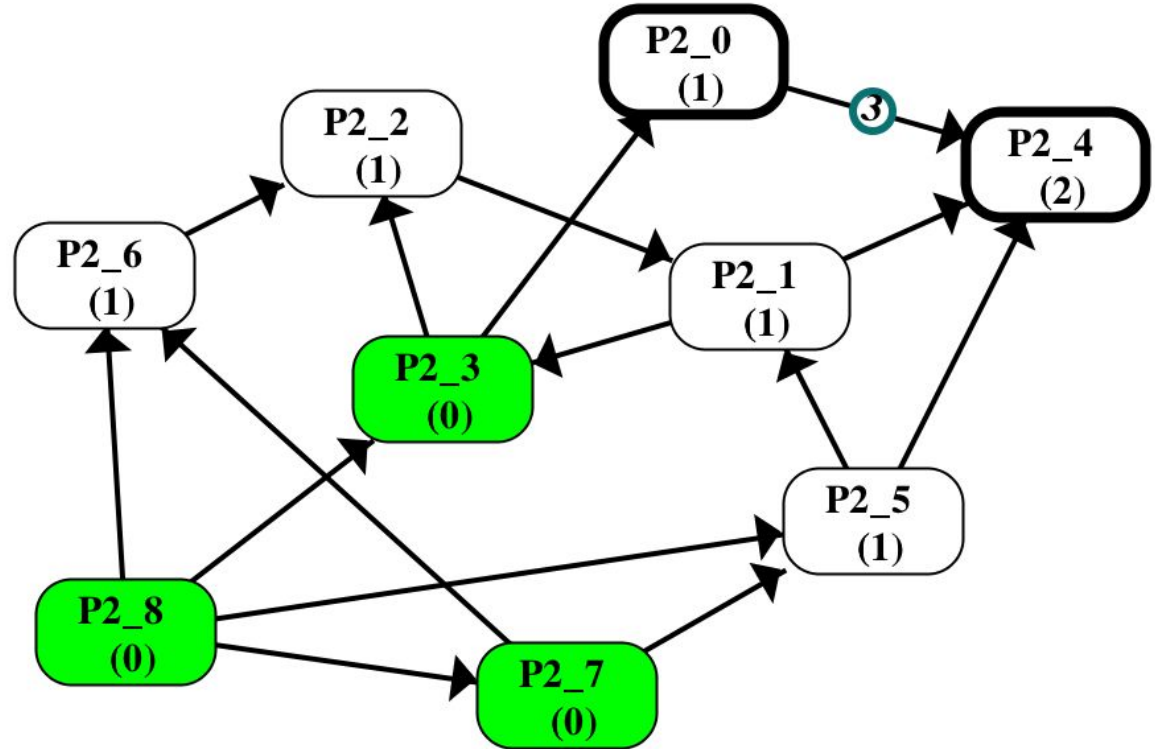
Local Network

Company 3



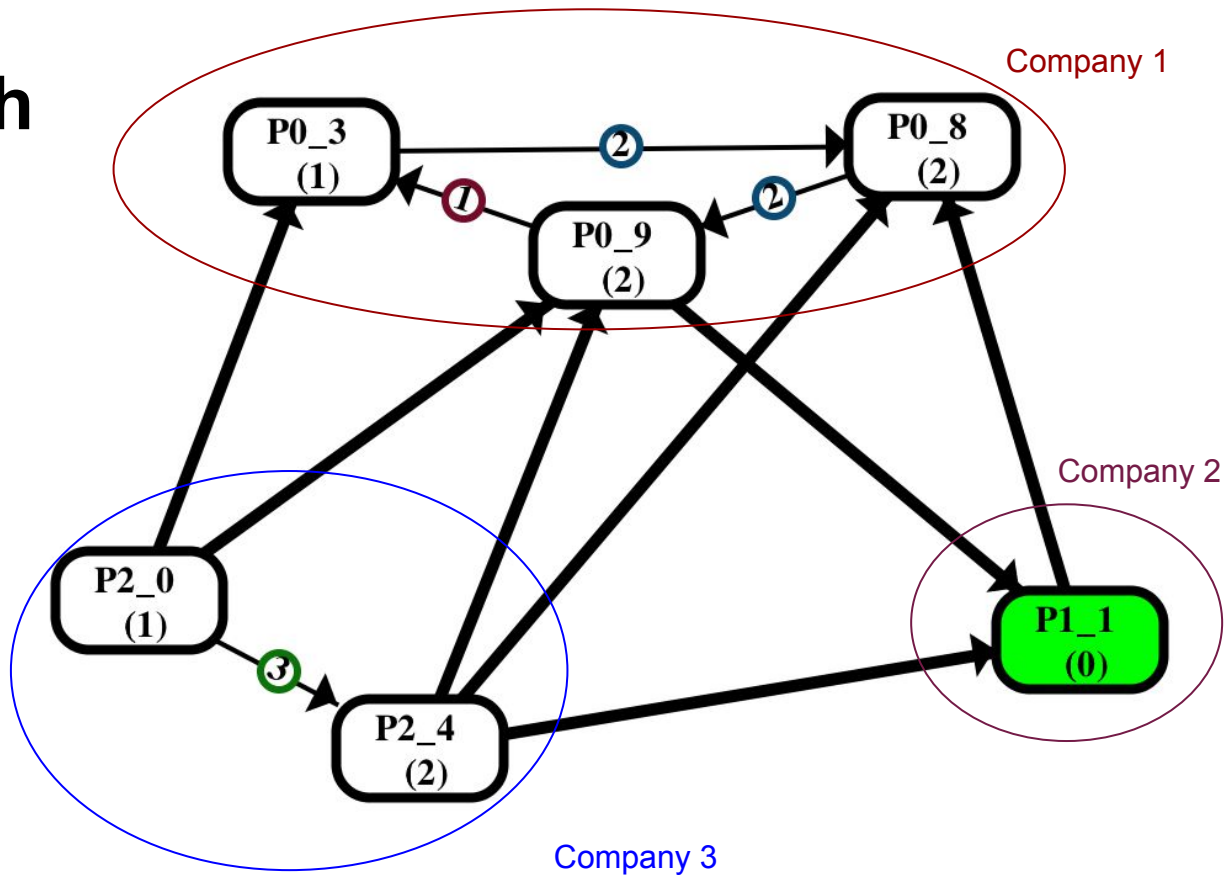
Local Computation Output

Company 3



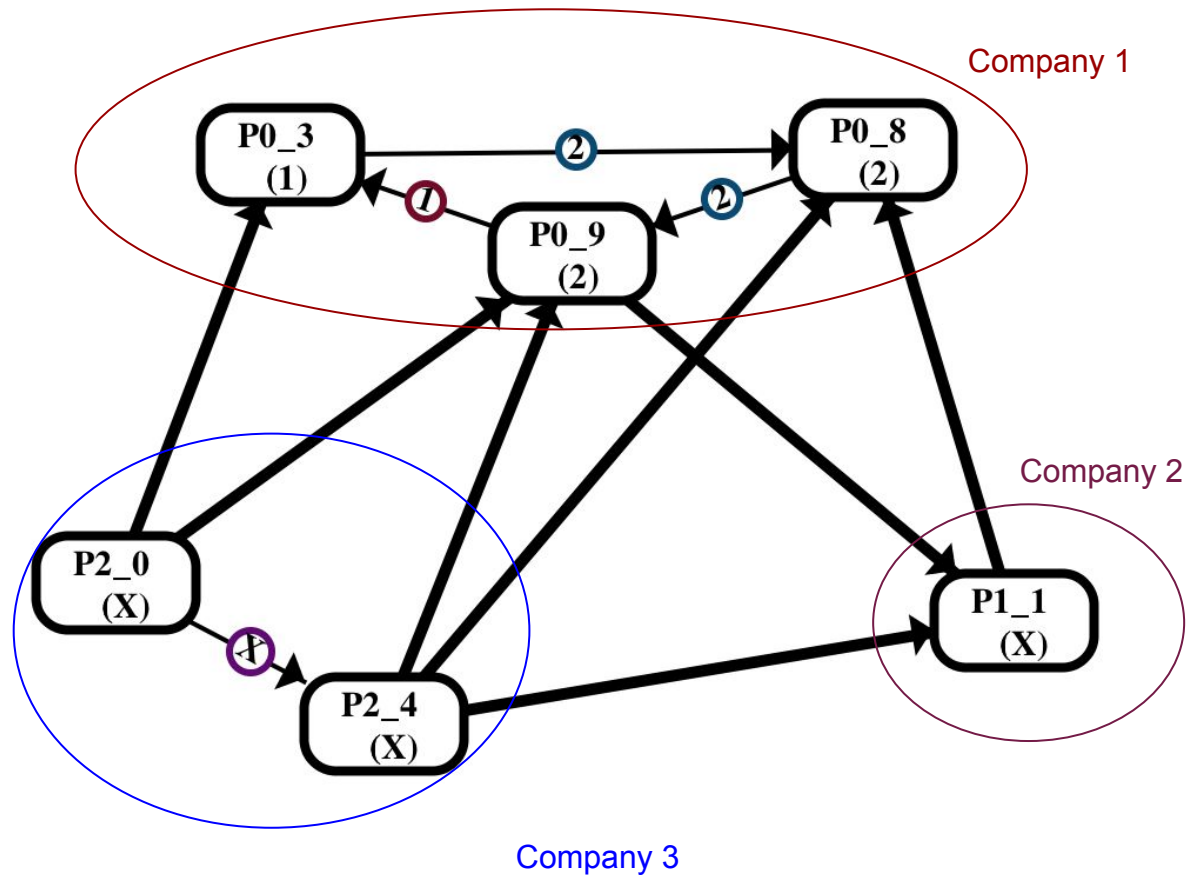
Gateways Graph

Global View



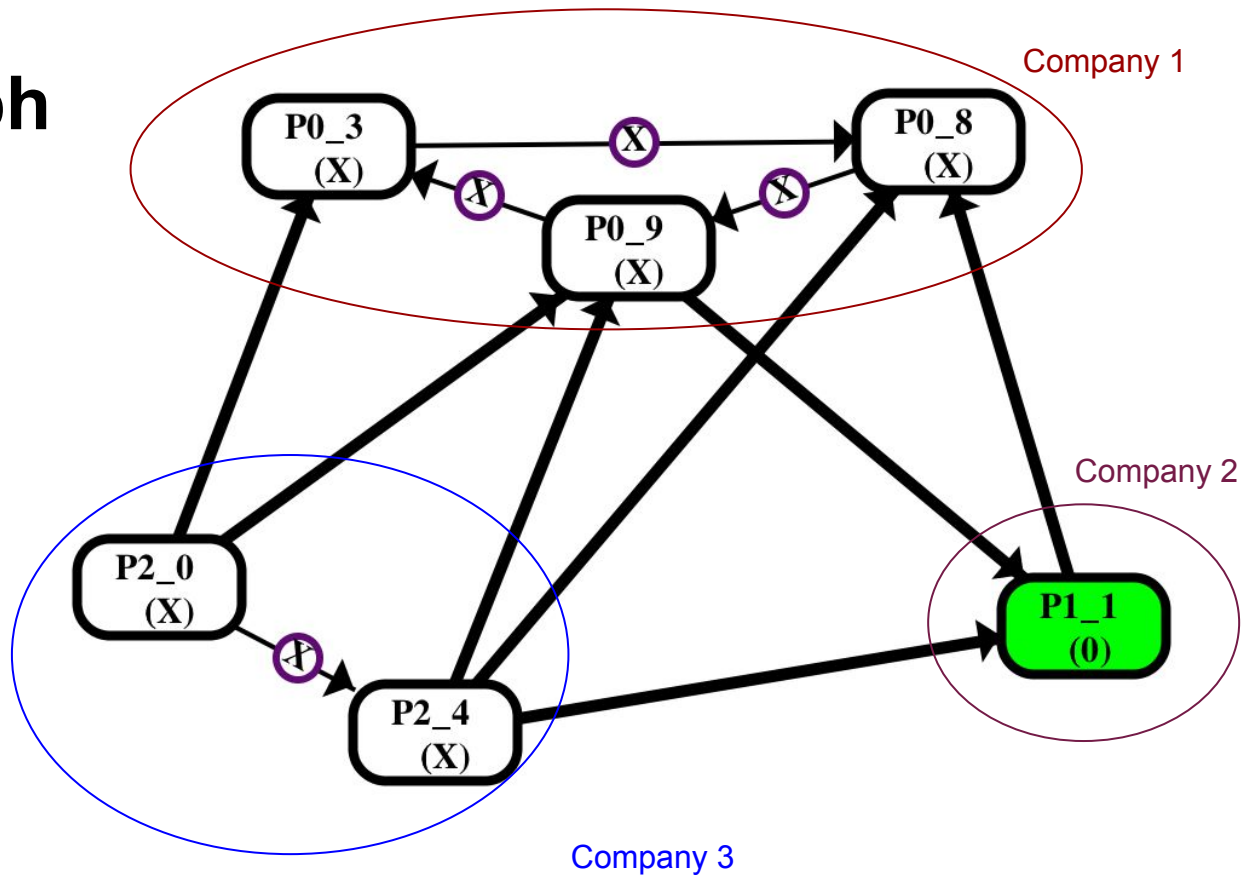
Gateways Graph

Company 1



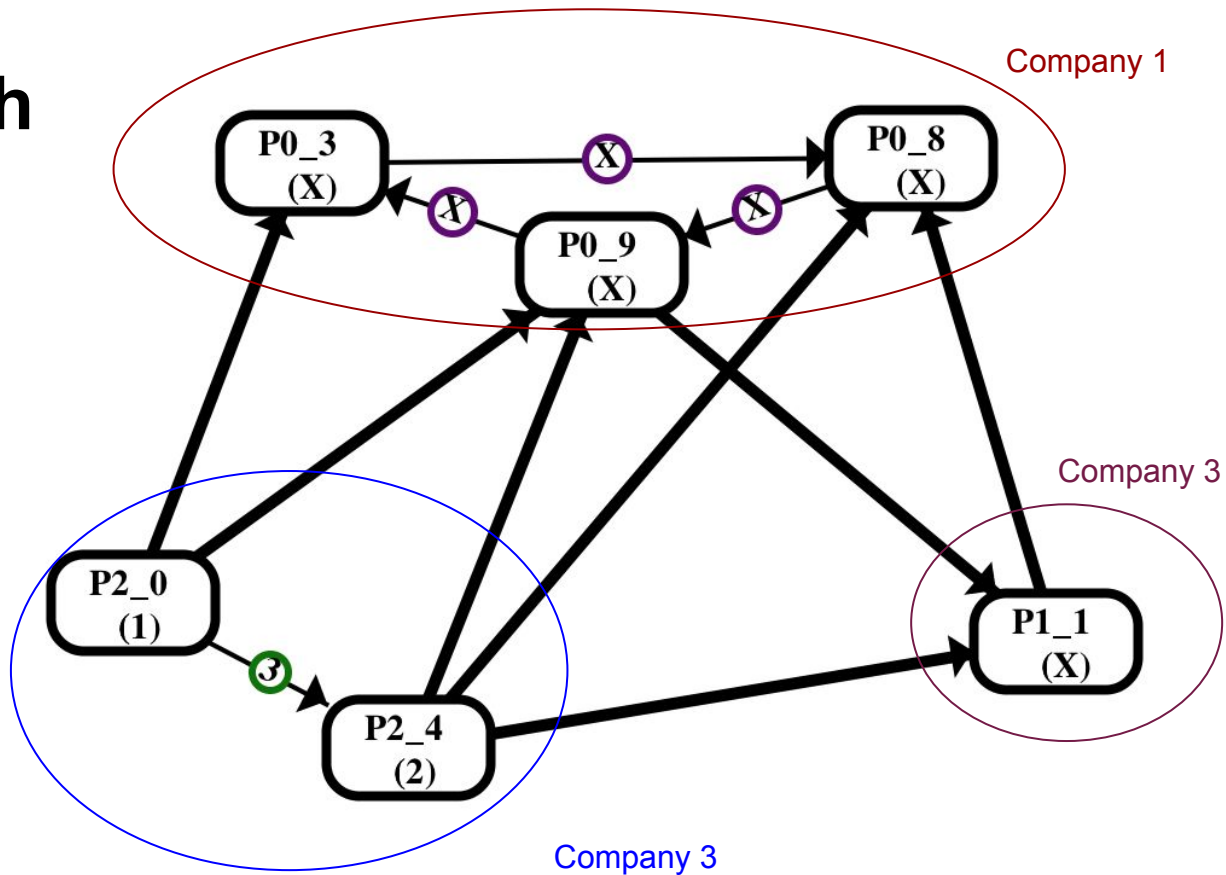
Gateways Graph

Company 2



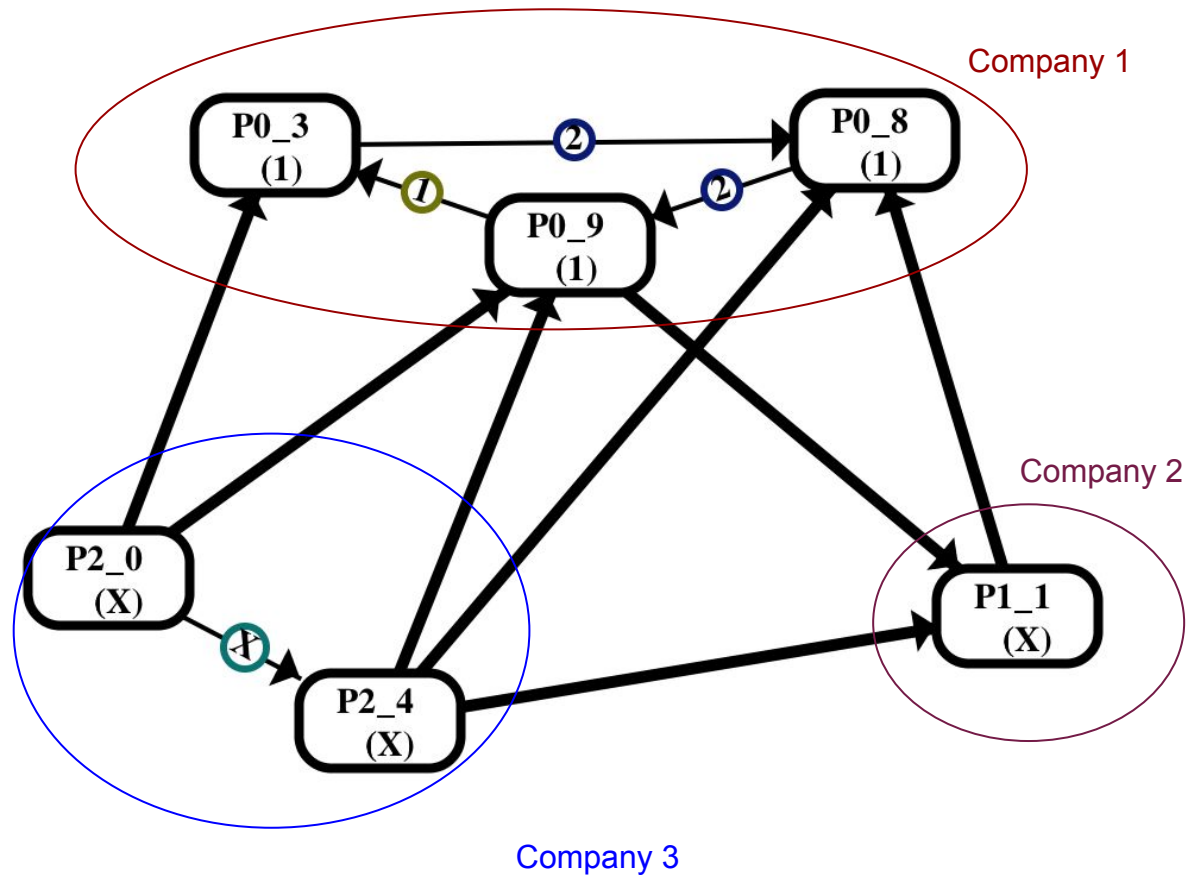
Gateways Graph

Company 3



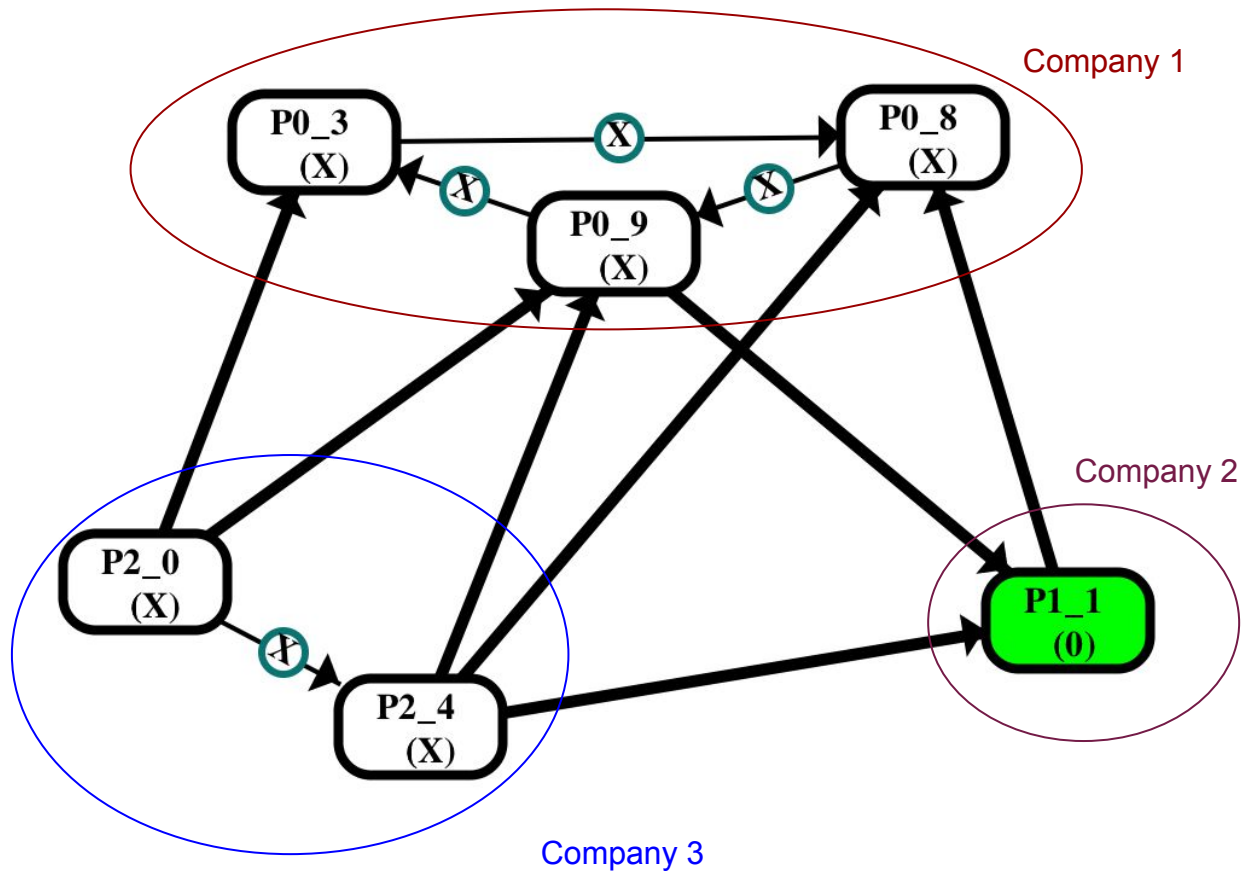
MPC Output

Company 1



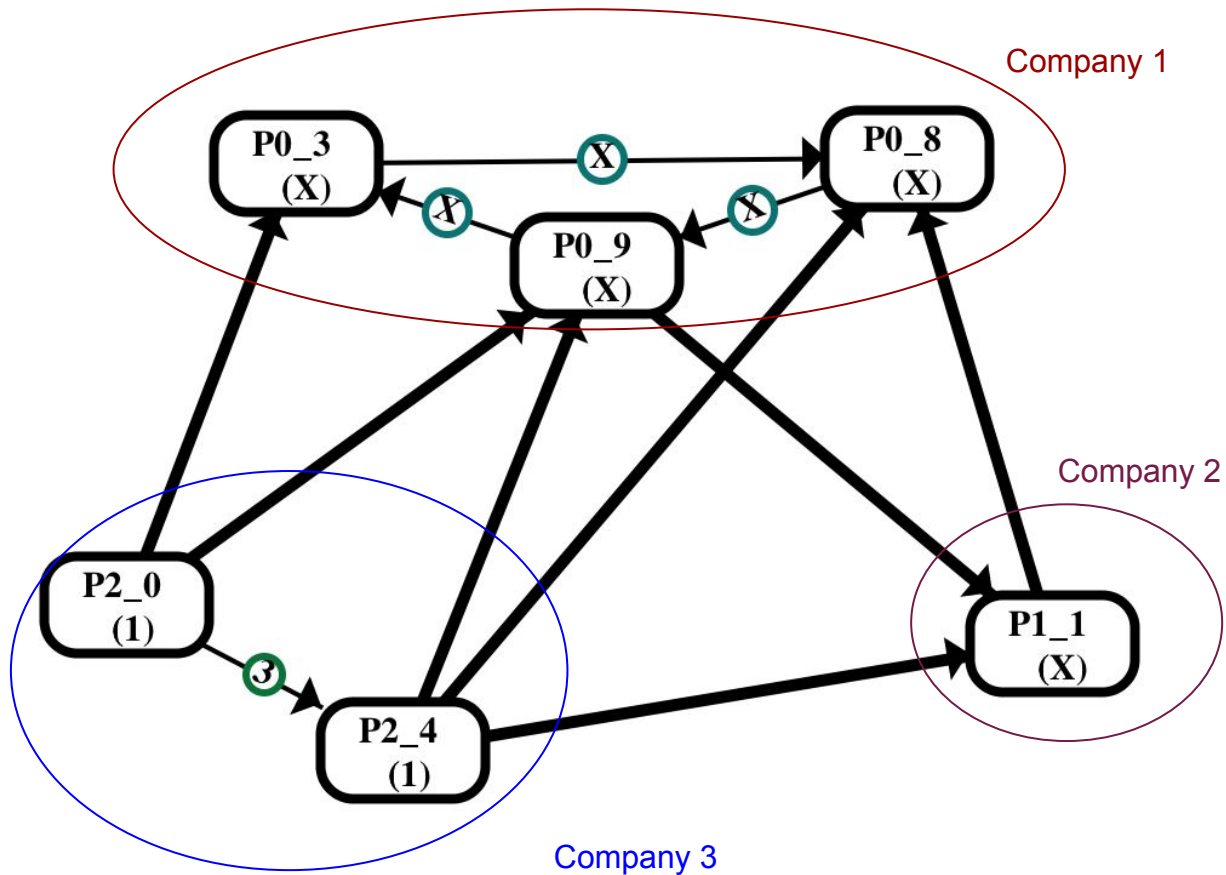
MPC Output

Company 2



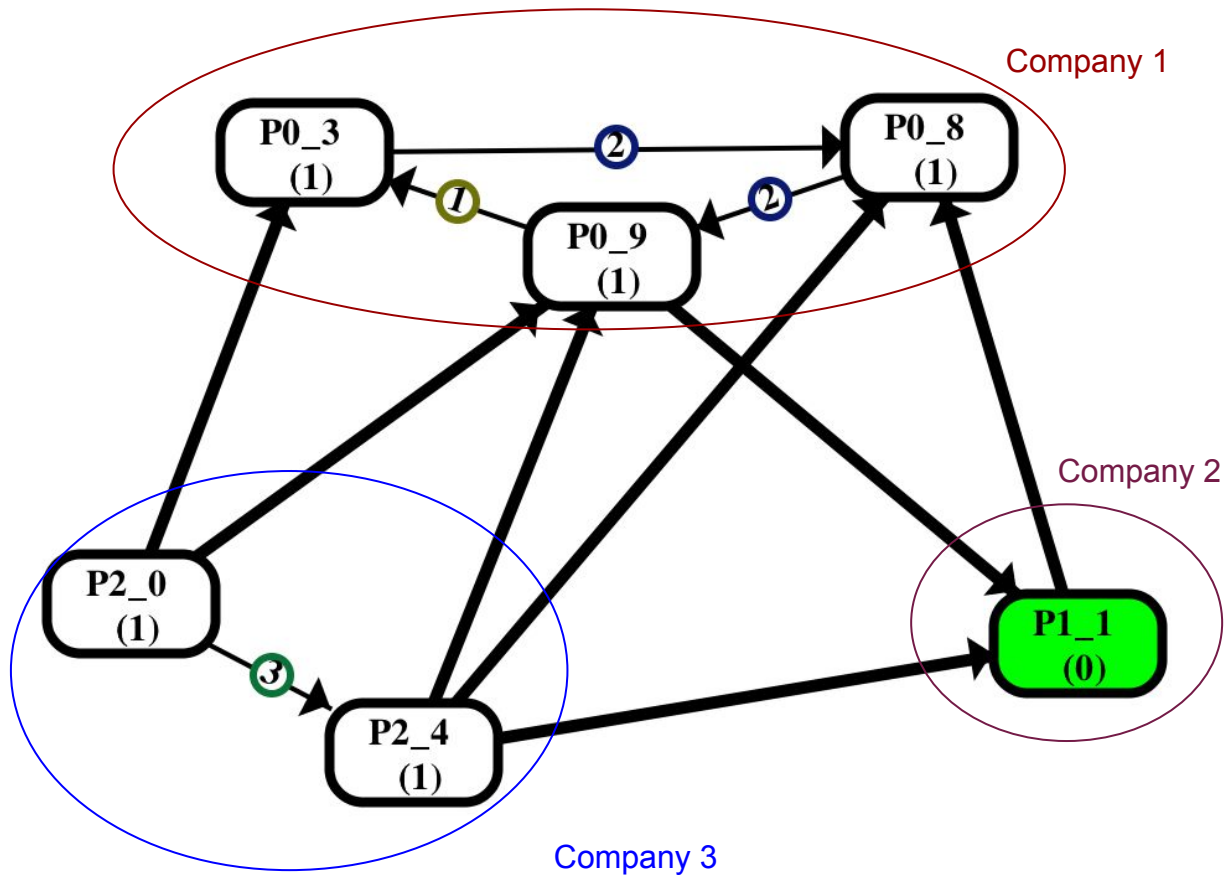
MPC Output

Company 3



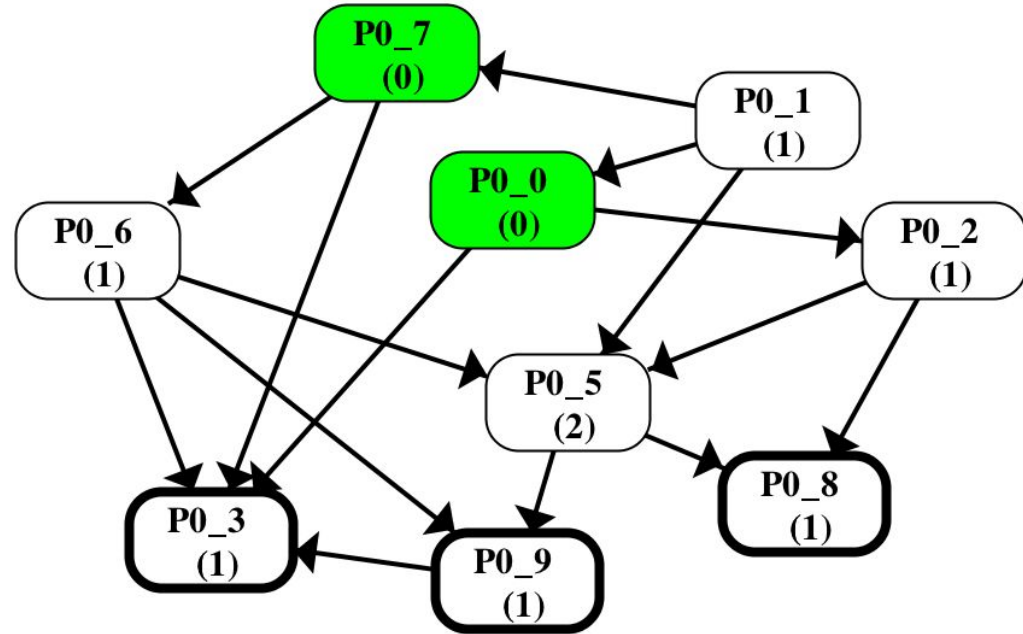
MPC Output

Global View



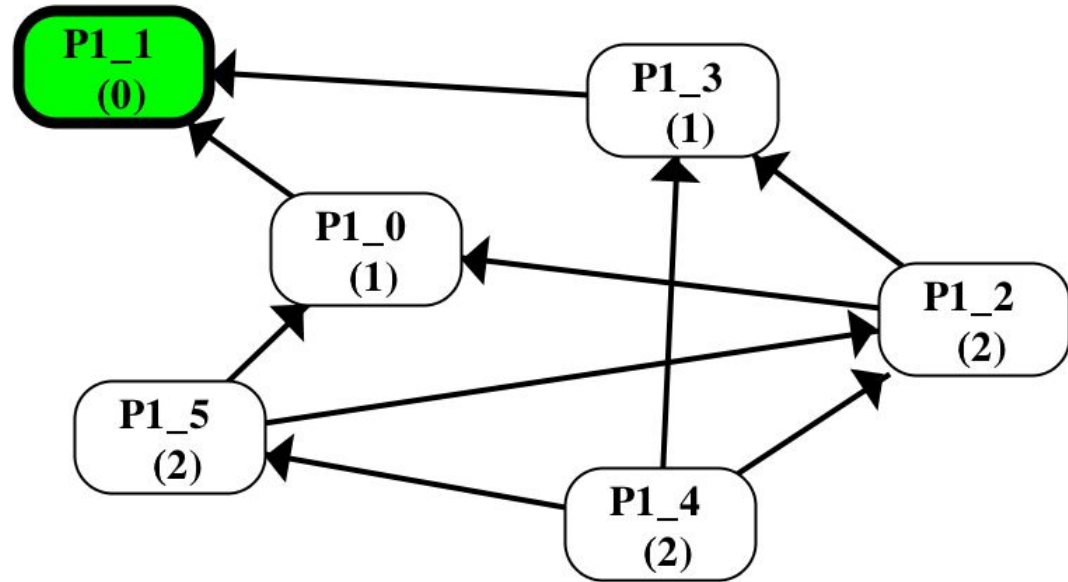
Final Output

Company 1



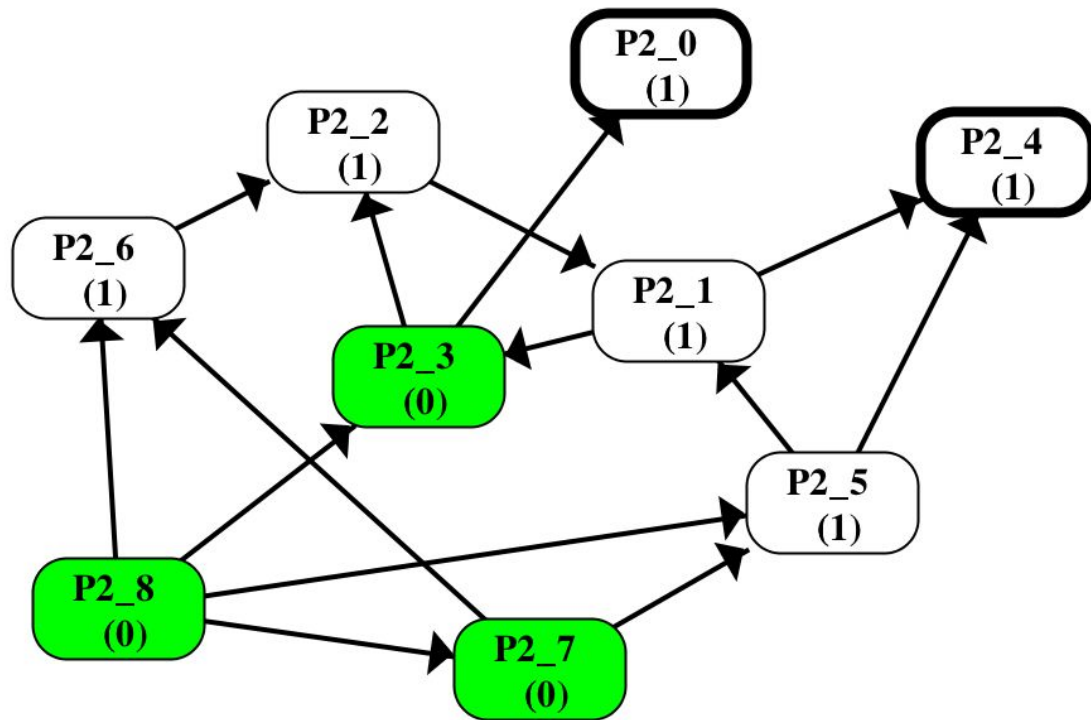
Final Output

Company 2



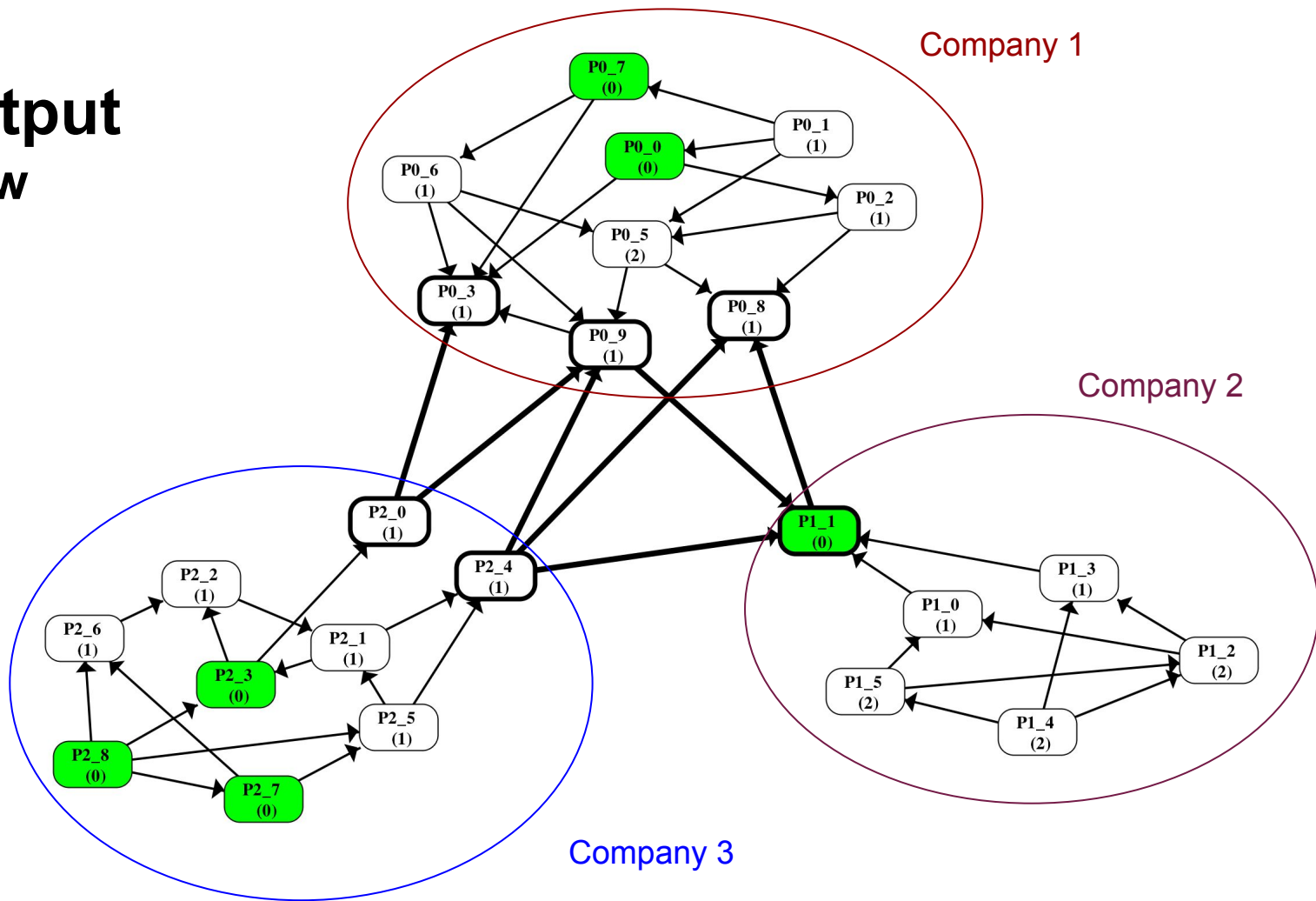
Final Output

Company 3



Final Output

Global View



Problem

- Not revealing any intermediate values
 - Can be done with viff, but with huge cost (min is expensive).

Solution

- Run algorithm with “symbolic” inputs.
- Unwraps iteration/execution yielding an expression.
- Secret share initial distances, evaluate expression in one shot.
- Only final answer is revealed.
- Expression can be optimized. It can be stored and re-used.

Expression

$P2_G2_1$

$$= \min(P2_G2_0, P1_G1_0+1, P2_G3_0+3, P2_G1_0+1)$$

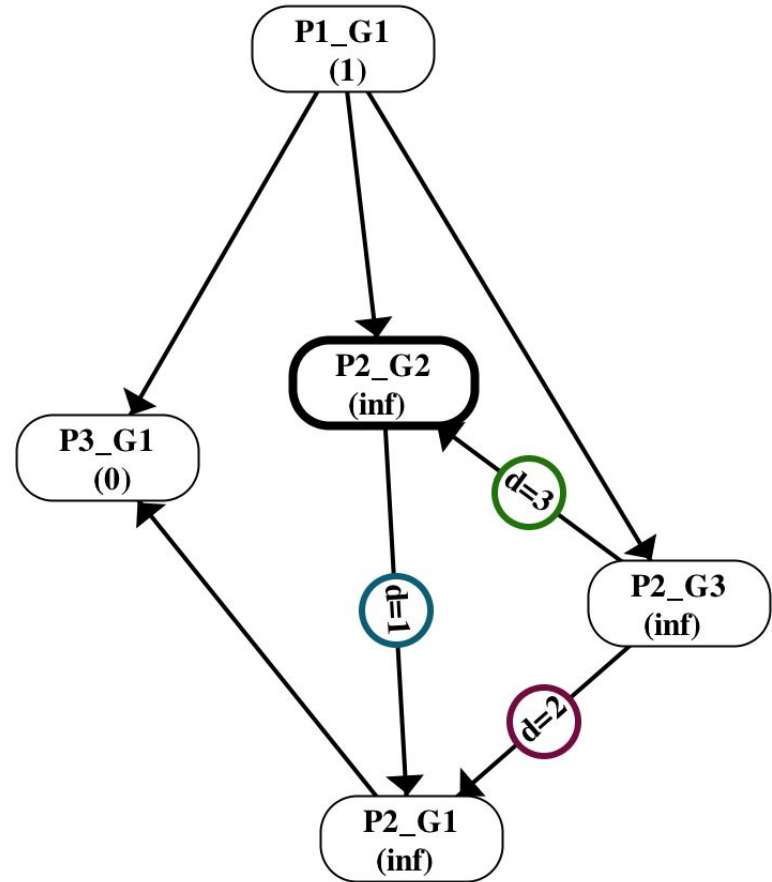
$$= \min(\text{inf}, 2, \text{inf}, \text{inf})$$

$P2_G2_2$

$$= \min(P2_G2_1, P1_G1_1+1, P2_G3_1+3, P2_G1_1+1)$$

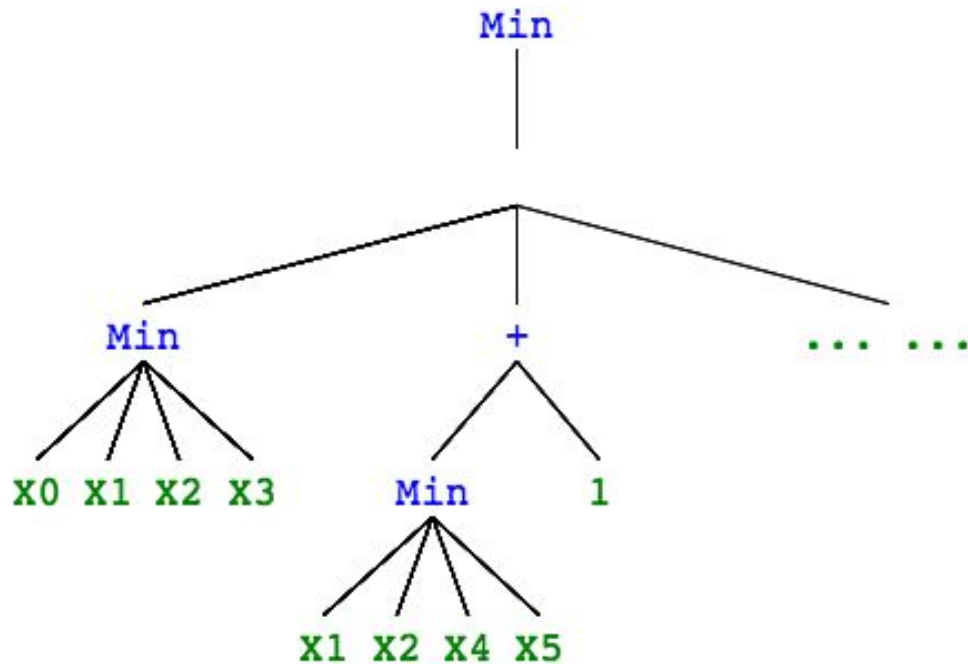
$$= \min(\min(\text{inf}, 2, \text{inf}, \text{inf}), \min(1, 1, \text{inf}, \text{inf})+1, \dots)$$

Note that this is a different example than the previously illustrated one

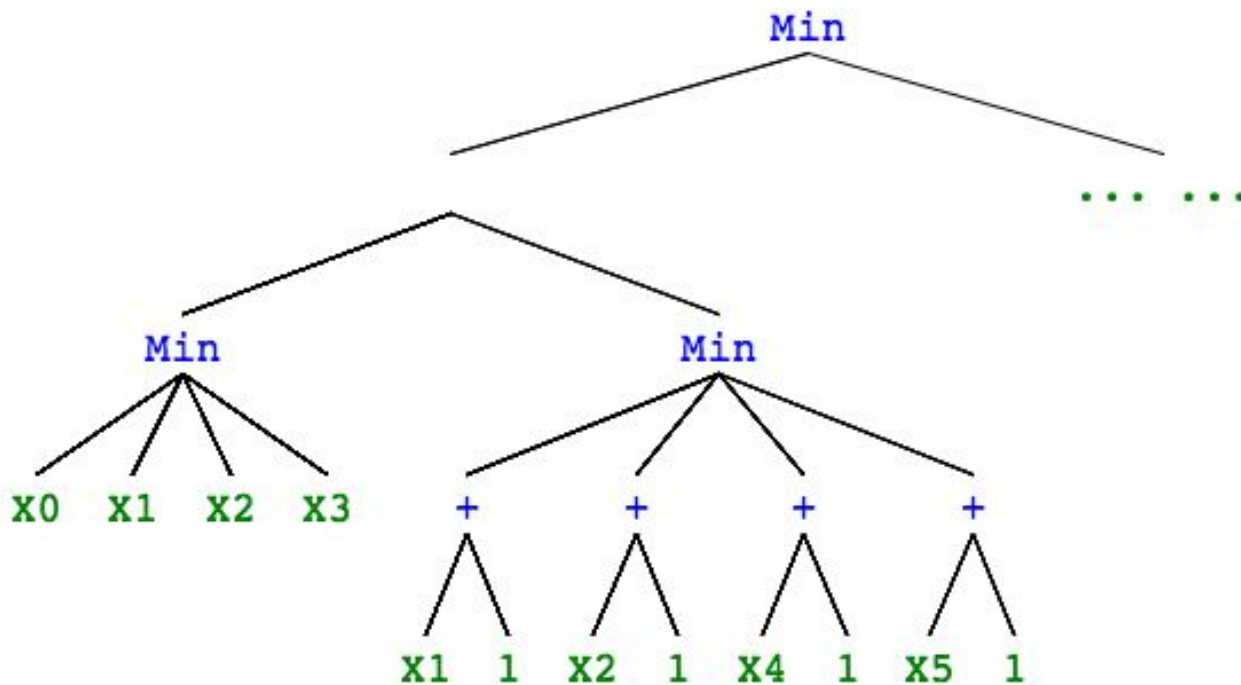


Expression Tree

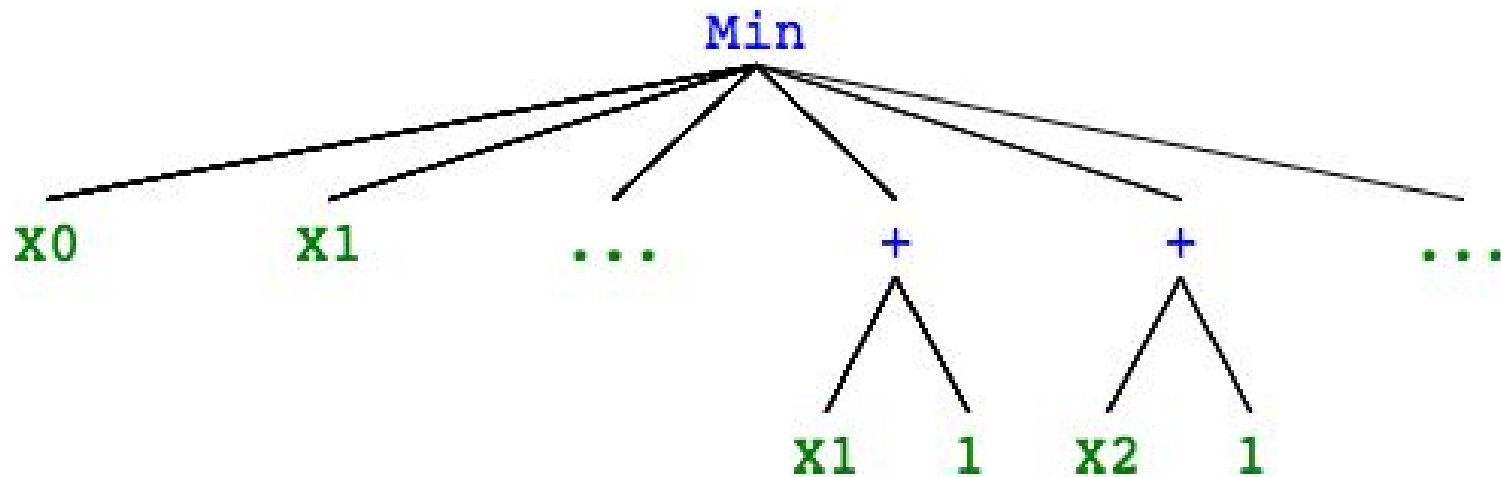
- Expressions as parse trees.
- Levels of + and Min.
- Optimizing the syntax tree is equivalent to simplifying the expression.
- Simplification utilizes distribution properties of + and Min.
- Simplification yields a single Min with compact additions.



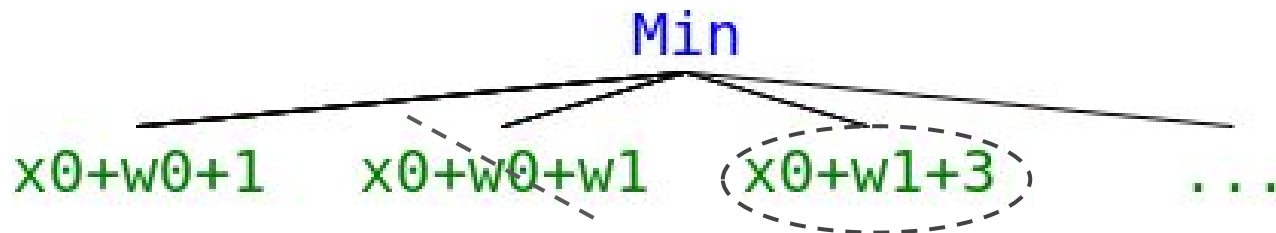
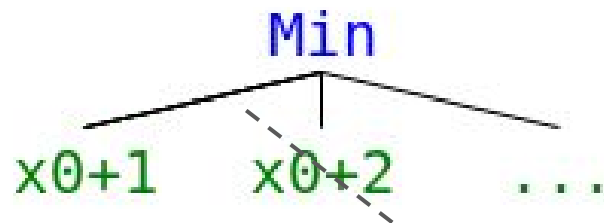
Add-Min Reduction



Min-Min Reduction



Early-Min Reduction



Library Modules

- Implemented as a Python Library.
- Operator overloading to construct Expressions.
- Important Modules
 - Expression : AddExpression, MinExpression, VarExpression ...
 - Simplifiers: BaseSimplifier, Reductions, Optimizing Simplifiers ...
 - Evaluators: BaseEvaluator (Direct), ViffEvaluator.
- Customize expression, simplifiers, and evaluators (inheritance).
- Code does not reveal MPC, Expressions, or Twisted.
- Example: Simplification: 0.12 (sec). Eval: 16(sec)

Pure Viff Solution: 24 (sec).

Future Extensions

- Provide builtin support for more expressions and simplifications.
- Apply the library/technique to more problems.
- Include conditionals, loops, and other python statements
- Speed-up evaluation by exploiting similar terms (memoization).
- SPDZ Evaluator ?



Thank You

Questions?