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Constraint Handling Rules -
Basic CHR programs and their
analysis

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Overview

Analysis of CHR programs regarding

- ▶ Logical reading and program correctness
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 - ▶ Actual worst-case complexity in CHR (refined semantics)
- ▶ Confluence
- ▶ Anytime and online algorithm property
- ▶ Concurrency and parallelism

Multiset transformation

- ▶ Programs consisting of essentially one constraint
- ▶ Constraint represents active data
- ▶ Pairs of constraints rewritten by single simplification rule
- ▶ Often possible: more compact notation with simpagation rule
- ▶ Simpagation rule removes one constraint, keeps (and updates) other

Minimum

Minimum program

```
min(N) \ min(M) <=> N=<M | true.
```

- ▶ Computes minimum of numbers given as $\text{min}(n_1), \text{min}(n_2), \dots, \text{min}(n_k)$
- ▶ Keeps removing larger values until only one value remains

Example computation

```
min(1), min(0), min(2), min(1)
```

```
min(0), min(2), min(1)
```

```
min(0), min(1)
```

```
min(0)
```

Logical reading (I)

- ▶ `min` constraints represent candidates for minimum
 - ▶ Actual minimum remains when calculation finished
 - ▶ Cannot be expressed straightforward in first-order logic
- ▶ First-order logic reading

$$\forall(N \leq M \rightarrow (\text{min}(N) \wedge \text{min}(M) \leftrightarrow \text{min}(N)))$$

Logically equivalent to

$$N \leq M \rightarrow (\text{min}(M) \leftarrow \text{min}(N))$$

- ▶ “Given a minimum, any larger value is also a minimum”

Logical reading (II)

- ▶ Linear logic reading

$$!\forall ((N \leq M) \multimap (min(N) \otimes min(M) \multimap min(N)))$$

- ▶ Reads as: Of course, consuming $min(N)$ and $min(M)$ where $(N \leq M)$ produces $min(N)$
- ▶ Properly reflects the dynamics of the minimum computation.

Correctness

Correctness by contradiction

- ▶ Minimum is not correctly computed
 - ▶ Case 1: more than one `min` constraint left
 - ▶ Case 2: remaining `min` constraint does not contain minimum
- ▶ Case 1: rule is still applicable
- ▶ Case 2: minimum must have been removed
 - ▶ Contradiction: rule always removes larger value

Termination and worst-case complexity

- ▶ Termination
 - ▶ Rule removes constraints, does not introduce new ones
 - ▶ Rule application in constant time (applies to every pair of `min` constraints)
 - ▶ Number of rule applications (derivation length) bounded by number of `min` constraints
- ▶ Worst-case time complexity
 - ▶ Given n `min` constraints
 - ▶ $O(n)$ under refined semantics (left-to-right, immediate reaction, one constraint will be removed)

Meta-complexity

- ▶ Abstract semantics: undetermined order of tried constraints and rules
- ▶ Meta-complexity theorem (MCT)

$$O\left(D \sum_i ((n + D)^{n_i} (O_{H_i} + O_{G_i}) + (O_{C_i} + O_{B_i}))\right),$$

(D derivation length, i ranges over rules, n_i number of head constraints in i th rule, costs O_{H_i} of head matching, O_{G_i} of guard checking, O_{C_i} of imposing built-in constraints of body, O_{B_i} of imposing CHR constraints of body)

- ▶ In this case

$$O(n(n^2(1 + 0) + (1 + 0))) = O(n^3).$$

- ▶ Highly over-estimates (applies to all two-head simpagation rules)

Confluence (I)

- ▶ Correctness implies result is single specific \min constraint
⇒ Program is confluent for ground queries (ground confluent)
- ▶ One rule, only overlaps with itself
- ▶ One nontrivial full overlap (all head constraints equated):
 $\min(A), \min(B), A=<B, B=<A.$
(equivalent to $\min(A), \min(A), A=B.$)
- ▶ Apply rule in given or reversed order
 - ▶ Both cases lead to $\min(A), A=B$ (hence rule removes duplicates)

Confluence (II)

- Four overlaps where one constraint shared

$\min(A), \min(B), \min(C), A=<B, B=<C.$

$\min(A), \min(B), \min(C), A=<B, B=<C.$

$\min(A), \min(B), \min(C), A=<B, A=<C.$

$\min(A), \min(B), \min(C), A=<B, C=<B.$

- First (and second) overlap leads to joinable critical pair

$$\begin{array}{c}
 \min(A), \min(B), \min(C), A=<B, B=<C \\
 / \qquad \qquad \qquad \backslash \\
 \min(A), \min(B), A=<B, B=<C \quad \min(A), \min(C), A=<B, B=<C \\
 \backslash \qquad \qquad \qquad / \\
 \min(A), A=<B, B=<C
 \end{array}$$

- Only smallest constraint $\min(A)$ is left

Confluence (III)

► Next overlap (similar)

$$\begin{array}{ccc}
 \min(A), \min(B), \min(C), A=<B, A=<C & & \\
 / & & \backslash \\
 \min(A), \min(B), A=<B, A=<C & \min(A), \min(C), A=<B, A=<C & \\
 \backslash & & / \\
 \min(A), A=<B, A=<C & &
 \end{array}$$

► Last overlap

$$\begin{array}{ccc}
 \min(A), \min(B), \min(C), A=<B, C=<B & & \\
 | & & | \\
 \min(A), \min(C), A=<B, C=<B & &
 \end{array}$$

- Cannot proceed until relationship between A and C known (but then common state is reached)

⇒ Program is confluent

Anytime algorithm property

- ▶ Anytime algorithm (approximation)
 - ▶ One can interrupt program at any time and restart on immediate result
 - ▶ On interrupt: subset of initial `min` constraints containing actual minimum
 - ⇒ interruption and restart possible
 - ▶ Intermediate results approximate final result
 - ▶ Set of possible minima gets smaller and smaller
- ⇒ Program is an anytime algorithm

Online algorithm property

- ▶ Online (incremental)
 - ▶ Possibility to add constraints while program is running
 - ▶ Additional `min` constraints can be added at any point
 - ▶ Immediately react with other constraints
 - ▶ Confluence guarantees same result, no matter when constraint is added

⇒ Program is incremental

Concurrency and parallelism (I)

- ▶ Program is well-behaved (terminating, confluent)
 - ⇒ parallelization easy
- ▶ Weak parallelism
 - ▶ Apply rule to different nonoverlapping parts of query
 - ▶ Rule can be applied to pairs of `min` constraints in parallel
 - ▶ Halves number of `min` constraints in each parallel computation step
 - ▶ $O(\log(n))$ on $n/2$ parallel processing units (processors)

Example computation

```

min(1), min(0),          min(2), min(1)
min(0),                min(1)
min(0)
  
```


Concurrency and parallelism (II)

- ▶ Strong parallelism
 - ▶ Apply rule to overlapping parts of query (fix one `min` constraint to be kept)
 - ▶ Linear complexity as in sequential execution (worst-case: with largest value fixed, no rule application possible)
- ▶ Cost (Time complexity times number of processors)
 - ▶ Parallel execution: $O(n \log(n))$
 - ▶ Sequential execution: $O(n)$

Boolean XOR

XOR program

```
xor(X), xor(X) <=> xor(0).  
xor(1) \ xor(0) <=> true.
```

- ▶ Implements Exclusive Or operation of propositional logic (0 means false, 1 means true)
- ▶ Query: multiset of `xor` constraints for input truth values (e.g. `xor(1)`, `xor(0)`, `xor(0)`, `xor(1)`)
- ▶ First rule: Identical inputs replaced by `xor(0)`
- ▶ Second rule: Remove `xor(0)` if there is `xor(1)`

Logical reading and correctness

- ▶ First-order logical reading
 - ▶ $xor(X) \leftrightarrow xor(0)$ (particularly $xor(1) \leftrightarrow xor(0)$)
 - ▶ Means all `xor` constraints are equivalent
 - ▶ Resort to linear logic reading
- ▶ Correctness
 - ▶ Map CHR conjunction to xor operation
 - ▶ Both associative, commutative, not idempotent
 - ▶ Each rule application computes one xor
 - ▶ One `xor` constraint left in the end

Termination and complexity

- ▶ Terminating
 - ▶ Each rule removes more constraints than it introduces
- ▶ Complexity
 - ▶ For each pair of constraints one rule application in constant time
 - ▶ Linear complexity under refined semantics
 - ▶ Cubic complexity under abstract semantics

Confluence (I)

XOR program

```
xor(X), xor(X) <=> xor(0).  
xor(1) \ xor(0) <=> true.
```

- ▶ **Overlap** `xor(X), xor(X)`
 - ▶ First rule fully with itself
 - ▶ Always leads to `xor(0)`
- ▶ **Overlap** `xor(X), xor(X), xor(X)`
 - ▶ First rule with itself
 - ▶ Always leads to `xor(0), xor(X)`

Confluence (II)

XOR program

```
xor(X), xor(X) <=> xor(0).  
xor(1) \ xor(0) <=> true.
```

- ▶ **Overlap** `xor(1), xor(1), xor(0)`
 - ▶ Occurs twice (first and second rule, second rule with itself)
 - ▶ Always leads to `xor(0)`
- ▶ **Overlap** `xor(1), xor(0), xor(0)`
 - ▶ Occurs twice (first and second rule, second rule with itself)
 - ▶ Always leads to `xor(1)`

⇒ Program is confluent

Remaining properties

- ▶ Anytime: fewer and fewer `xor` constraints, result not necessarily contained $(\text{xor}(1), \text{xor}(1))$
- ▶ Online: `xor` constraints can be added at any point
- ▶ Rules applicable in parallel (as for `min`) $\Rightarrow O(n \log(n))$

Greatest common divisor

GCD program

```

gcd(0) <=> true.
sub @ gcd(N) \ gcd(M) <=> 0<N,N=<M | gcd(M-N) .
mod @ gcd(N) \ gcd(M) <=> 0<N,N=<M | gcd(M mod N) .

```

- ▶ Either `sub` or `mod` rule can be used

Example computation (`sub`)

```

gcd(7), gcd(12)
gcd(7), gcd(5)
gcd(5), gcd(2)
gcd(2), gcd(3)
gcd(2), gcd(1)
gcd(1), gcd(1)
gcd(1), gcd(0)
gcd(1)

```

Example computation (`mod`)

```

gcd(7), gcd(12)
gcd(7), gcd(5)
gcd(5), gcd(2)
gcd(2), gcd(1)
gcd(1), gcd(0)
gcd(1)

```


Logical Reading

- ▶ First-order logical reading

$$\text{gcd}(0) \leftrightarrow \text{true}$$

$$0 < N \wedge N = < M \rightarrow (\text{gcd}(N) \wedge \text{gcd}(M) \leftrightarrow \text{gcd}(N) \wedge \text{gcd}(M - N))$$

Latter is equivalent to

$$0 < N \wedge 0 = < M \rightarrow (\text{gcd}(N) \wedge \text{gcd}(M + N) \leftrightarrow \text{gcd}(N) \wedge \text{gcd}(M))$$

- ▶ Correct, but does not characterize gcd, only all its multiples
- ▶ Linear-logic semantics reflects dynamics of computation properly

Correctness

- ▶ All divisors d preserved under rule application
- ▶ Computation produces smaller and smaller values
 - ▶ $N = Ad, M = Bd$
 - ▶ From logical reading

$$0 < Ad \wedge Ad = < Bd$$

$$\rightarrow (\gcd(Ad) \wedge \gcd(Bd) \leftrightarrow \gcd(Ad) \wedge \gcd(Bd - Ad))$$

- ▶ $\gcd(Bd - Ad)$ is equivalent to $\gcd((B - A)d)$

\Rightarrow Divisor d preserved during computation

- ▶ Computation continues until $M = N = \gcd$
- ▶ Rule is applied a last time
- ▶ $\gcd(0)$ is removed leaving only actual \gcd

Termination and complexity

▶ Termination

- ▶ Guard condition ensure new value smaller than removed M
- ▶ New value cannot become negative

▶ Complexity

- ▶ Rules applicable in constant time to any gcd pair
- ▶ Two gcd constraints
 - ▶ sub : complexity linear in larger number
 - ▶ mod : complexity logarithmic in larger number
- ▶ More than two gcd constraints: consider all numbers
 - ▶ sub linear in sum of numbers
 - ▶ mod logarithmic in product of numbers

Confluence

- ▶ GCD program is ground confluent (unique result for given values)
- ▶ Not confluent in general:
 - ▶ Overlaps analogous to `min`
 - ▶ Difference: rule not only removes constraints but also adds
 - ▶ Nonjoinable critical pair (cp)

$$\begin{array}{c}
 \text{gcd}(A), \text{gcd}(B), \text{gcd}(C), 0 < A, A = < B, 0 < C, C = < B \\
 | \qquad \qquad \qquad \qquad \qquad \qquad \qquad \qquad | \\
 \text{gcd}(A), \text{gcd}(B-A), \text{gcd}(C), 0 < A, A = < B, 0 < C, C = < B \quad | \\
 \qquad \qquad \qquad \qquad \qquad \qquad \qquad \qquad \qquad \qquad | \\
 \text{gcd}(A), \text{gcd}(B-C), \text{gcd}(C), 0 < A, A = < B, 0 < C, C = < B
 \end{array}$$

- ▶ Computation cannot proceed until relationship of A , B , and C is known

Remaining properties

- ▶ Anytime: fewer and fewer gcd constraints with smaller and smaller numbers (result not necessarily contained)
- ▶ Online: additional gcd constraints can be added anytime
- ▶ Complexity of parallel execution not better nor worse than sequential (since $O(\max(a, b)) = O(a + b)$)
- ▶ But gcd 's may get smaller more quickly
- ▶ In practice: super-linear speed up with parallel CHR implementation in Haskell

Prime sieve

Prime sieve program

```
sift @ prime(I) \ prime(J) <=> J mod I ::= 0 | true.
```

- ▶ Removes multiples in given set until only prime numbers left
- ▶ Query: prime candidates from 2 upto N
(`prime(2)`, `prime(3)`, `prime(4)`, ..., `prime(N)`)

Example computation

```
prime(7), prime(6), prime(5), prime(4), prime(3), prime(2)  
prime(7), prime(5), prime(4), prime(3), prime(2)  
prime(7), prime(5), prime(3), prime(2)
```

Logical reading and correctness

- ▶ First-order logical reading

$$\forall((M \bmod N = 0) \rightarrow (\text{prime}(M) \wedge \text{prime}(N) \leftrightarrow \text{prime}(N)))$$

- ▶ Means a number is prime if it is a multiple of another prime number
 - ▶ Linear logic reading reflects dynamics of filtering correctly
- ▶ Correctness
 - ▶ Program confluent \Rightarrow result always the same
 - ▶ All composite numbers removed (with correct query)
 - ▶ Primes not removed (only multiple of 1, not included)

Termination and complexity

- ▶ Termination
 - ▶ Rule only removes constraints
- ▶ Complexity
 - ▶ Rule not applicable to all pairs of numbers
 - ▶ Thus complexity quadratic in number of constraints (refined semantics)
 - ▶ Runtime can be improved by starting from lower numbers

Confluence

► Program is confluent

- Reason: transitivity of divisibility

► $I|J$ and $J|K \Rightarrow I|K$

- Overlaps and joinability analogous to `min`

`prime(A), prime(B), A|B, B|A`

`prime(A), prime(B), prime(C), A|B, B|C`

`prime(A), prime(B), prime(C), A|B, A|C`

`prime(A), prime(B), prime(C), A|B, C|B`

- First three overlaps lead to joinable critical pair

- Last overlap also:

`prime(A), prime(B), prime(C), A|B, C|B`

$$\begin{array}{ccc} | & & | \\ \text{prime(A), prime(C), A|B, C|B} \end{array}$$

Other properties

- ▶ Anytime and online properties as for \min
- ▶ `sift` does not hold for all pairs
 - ▶ All $O(n^2)$ pairs have to be tried in $O(n)$ rounds
⇒ some scheduling needed
- ▶ Strong parallelism
 - ▶ Fix one prime constraint for first head constraint
 - ▶ Search for prime constraint matching second head constraint
 - ▶ Needs $O(n)$ rounds
- ▶ Cost same as for sequential execution (quadratic)
- ▶ Linear time: maximal, linear parallel speed-up

Exchange sort

Exchange sort program

$a(I, V), a(J, W) \Leftarrow I > J, V < W \mid a(J, V), a(I, W).$

- ▶ Exchanges values that are in the wrong order
- ▶ Query: array of values A_i ($a(1, A_1), \dots, a(n, A_n)$)

Example computation

$a(0, 1), a(1, 7), a(2, 5), \underline{a(3, 9)}, \underline{a(4, 2)}$

$a(0, 1), a(1, 5), \underline{a(2, 7)}, \underline{a(3, 2)}, a(4, 9)$

$a(0, 1), \underline{a(1, 5)}, \underline{a(2, 2)}, a(3, 7), a(4, 9)$

$a(0, 1), a(1, 2), a(2, 5), a(3, 7), a(4, 9)$

Logical reading and correctness

- ▶ First-order logical reading

$$I > J \wedge V < W \rightarrow (a(I, V) \wedge a(J, W) \leftrightarrow a(J, V) \wedge a(I, W))$$

- ▶ Means all arrays with same set of values are equivalent
- ▶ Resort to linear logic reading

- ▶ Correctness

- ▶ Sorted: for each $(a(I, V), a(J, W))$ with $I > J$ it holds that $V \geq W$
- ▶ If condition $V \geq W$ does not hold, rule is applicable
 - ⇒ condition holds after application
 - ⇒ if rule not applicable, array must be sorted

Termination

- ▶ Rule application cannot introduce more wrong than right orderings
 - ▶ Guard: counter in each array entry
 - ▶ Counts how many values with larger index are smaller
 - ▶ On exchange:
 - ▶ Counter of smaller value increases
 - ▶ Counter of larger value decreases by same number +1
 - ▶ Counter of values in between can only decrease
- ⇒ Sum of counters decrease with each rule application

Complexity

- ▶ Derivation length quadratic in number of constraints (cf. counter)
- ▶ Two head constraints: MCT gives overestimated complexity

$$O(n^2((n^2)^2(1+1) + (0+1))) = O(n^6)$$

- ▶ Fix one value (refined semantics): Each try costs $O(n)$
 - ▶ Rule can be applied in constant time once pair found
- ▶ At most $O(n^2)$ applications \Rightarrow actual worst-case complexity $O(n^3)$

Confluence (I)

- ▶ Program is ground confluent by correctness (unique result)
- ▶ Not confluent in general
- ▶ First critical pair is joinable

$$\begin{array}{c}
 a(I, V), a(J, W), a(K, U), I > J, V < W, J > K, W < U \\
 \quad / \quad I, J \qquad \qquad \qquad \qquad \qquad \qquad | \quad J, K \\
 \quad / \qquad \qquad a(I, V), a(K, W), a(J, U), I > J, V < W, J > K, W < U \\
 \quad / \qquad \qquad \qquad \qquad \qquad \qquad \qquad \qquad \qquad | \\
 a(J, V), a(I, W), a(K, U), I > J, V < W, J > K, W < U \quad | \\
 \quad | \quad I, K \qquad \qquad \qquad \qquad \qquad \qquad | \quad I, K \\
 \quad | \quad a(K, V), a(I, W), a(J, U), I > J, V < W, J > K, W < U \\
 \quad | \qquad \qquad \qquad \qquad \qquad \qquad \qquad \qquad \qquad | \\
 a(J, V), a(K, W), a(I, U), I > J, V < W, J > K, W < U \quad | \\
 \quad \backslash \quad J, K \qquad \qquad \qquad \qquad \qquad \qquad | \quad I, J \\
 \quad \backslash \qquad \qquad \qquad \qquad \qquad \qquad \qquad \qquad \qquad | \\
 a(K, V), a(J, W), a(I, U), I > J, V < W, J > K, W < U
 \end{array}$$

Confluence (II)

► Two nonjoinable critical pairs

$$\begin{array}{l}
 a(I, V), a(J, W), a(K, U), I > J, V < W, I > K, V < U \\
 / I, J \qquad \qquad \qquad | \\
 a(J, V), a(I, W), a(K, U), I > J, V < W, I > K, V < U \quad | \\
 \qquad \qquad \qquad \qquad \qquad \qquad \qquad \qquad \qquad | I, K \\
 a(K, V), a(J, W), a(I, U), I > J, V < W, I > K, V < U
 \end{array}$$

- Only joinable when relationship between J and K as well as W and U known

► Analogous situation for

$$a(I, V), a(J, W), a(K, U), I > K, V < U, J > K, W < U$$

Remaining properties

- ▶ Number of wrongly ordered pairs decreases over time
- ▶ Additional array entries can be added at any point
- ▶ Rule not applicable to arbitrary pairs of constraints
⇒ only weak parallelism possible
 - ▶ Associate each array entry with a processor
 - ▶ Try all pairs in $O(n)$ (macro-step)
 - ▶ Each entry reacts with at most $O(n)$ other entries
 - ▶ Overall $O(n^2)$ rule applications
 - ▶ All rule applications can be performed in $O(n)$ macro-steps ⇒
Complexity quadratic, cost cubic

Square root

Square root program

```
sqrt (X,G) <=> abs (G*G/X-1)>eps | sqrt (X, (G+X/G)/2) .
```

- ▶ Rule implements Newton's method
- ▶ `sqrt (X, G)` : square root of `X` is approximated by `G`
- ▶ `eps` is greater but close to 0
- ▶ Start with positive numbers `X` and `G`

Logical reading and termination

- ▶ Logical reading

$$abs(G * G/X - 1) > \epsilon \rightarrow (sqrt(X, G) \leftrightarrow sqrt(X, (G + X/G)/2))$$

- ▶ Means that any value is an approximation of \sqrt{X}
- ▶ Resort to linear logic reading

- ▶ Termination

- ▶ After first rule application $G \geq \sqrt{X}$
- ▶ If $G = \sqrt{X}$ rule not applicable
- ▶ Otherwise rule applicable, second argument will decrease

Remaining properties

- ▶ Confluence, anytime, online algorithm
 - ▶ Hold trivially (single rule with single head constraint)
- ▶ Concurrency, parallelism
 - ▶ Several constraints can run independently in parallel

Maximum

Maximum program

$\text{max}(X, Y, Z) \Leftarrow X \leq Y \mid Z = Y.$

$\text{max}(X, Y, Z) \Leftarrow Y \leq X \mid Z = X.$

- ▶ $\text{max}(X, Y, Z)$ means Z is the maximum of X and Y
- ▶ \leq and $<$ built-ins

Example computation

$\text{max}(1, 2, M)$: first rule applicable, reduces to $M=2$

$\text{max}(1, 2, 3)$: fails because of built-in $3=2$

$\text{max}(1, 1, M)$: both rules applicable, reduces to $M=1$

Logical reading and correctness

- ▶ First-order logical reading is

$$X \leq Y \rightarrow (\max(X, Y, Z) \leftrightarrow Z = Y)$$

$$Y \leq X \rightarrow (\max(X, Y, Z) \leftrightarrow Z = X)$$

- ▶ Logical consequences of the definition of \max

$$\max(X, Y, Z) \leftrightarrow (X \leq Y \wedge Z = Y \vee Y \leq X \wedge Z = X)$$

- ▶ This shows logical correctness

Termination and complexity

- ▶ One constraint removed in each step
 - ⇒ At most n (number of constraints) derivation steps
 - ▶ In each step at most n constraints checked against rules
 - ▶ Checking or establishing syntactic equality in constant time
 - ▶ Matching constraint against rule in quasi-constant time
 - ▶ Rule application in quasi-constant time
- ⇒ Worst-case complexity slightly worse than $O(n^2)$
- ▶ Same complexity is obtained using MCT

Remaining properties

▶ Confluence

- ▶ Only overlap is $\max(X, Y, Z) \wedge X \leq Y \wedge Y \leq X$
- ▶ Leads to critical pair

$$(Y = Z \wedge X \leq Y \wedge Y \leq X, X = Z \wedge X \leq Y \wedge Y \leq X)$$

- ▶ Both states equivalent to $X = Y \wedge Y = Z$
- ▶ Anytime, online algorithm
 - ▶ Hold trivially (single-headed simplification rule)
- ▶ Concurrency, parallelism
 - ▶ \max constraint may have to wait for result of other constraint
(e.g. $\max(X, Y, Z)$, $\max(Y, Z, W)$)

Fibonacci numbers

Fibonacci program

```
f0 @ fib(0,M) <=> M=1.  
f1 @ fib(1,M) <=> M=1.  
fn @ fib(N,M) <=> N>=2 |  
    fib(N-1,M1), fib(N-2,M2), M is M1+M2.
```

- ▶ `fib(N,M)` holds if `M` is `N`th Fibonacci number

Example computations

Query `fib(8,A)` yields `A=34`

Query `fib(12,233)` succeeds

Query `fib(11,233)` fails

Query `fib(N,233)` delays

Logical reading and correctness

- ▶ First-order logical reading

$$\mathit{fib}(0, M) \leftrightarrow M = 1$$

$$\mathit{fib}(1, M) \leftrightarrow M = 1$$

$$N \geq 2 \rightarrow (\mathit{fib}(N, M) \leftrightarrow \mathit{fib}(N - 1, M1) \wedge \mathit{fib}(N - 2, M2) \wedge M = M1 + M2)$$

- ▶ Shows correctness (coincides with mathematical definition)

Termination and complexity

- ▶ Program terminates
 - ▶ First argument of `fib` decreases in each call
 - ▶ Call only possible with positive first argument
- ▶ Ranking gives upper bound on derivation length

$$\text{rank}(\text{fib}(n, m)) = 2^n.$$

- ▶ Expected exponential complexity $O(2^n)$
- ▶ If first argument unknown complexity may increase (depending on wake-up policy)
- ▶ MCT reflects this and gives $O(4^n)$

Remaining properties

- ▶ Confluence:
 - ▶ No overlaps (single-headed simplification rules whose heads and guards exclude each other)
- ▶ Anytime, online algorithm, and concurrency
 - ▶ Hold trivially (single-headed simplification rule)

Fibonacci numbers (memorization version)

Fibonacci program with memorization

```

mem @ fib(N,M1) \ fib(N,M2) <=> M1=M2.

f0 @ fib(0,M) ==> M=1.
f1 @ fib(1,M) ==> M=1.
fn @ fib(N,M) ==> N>=2 |
    fib(N-1,M1), fib(N-2,M2), M is M1+M2.
  
```

Example computations

Query `fib(8,A)` returns all Fibonacci numbers up to 8:

```
fib(0,1), fib(1,1), fib(2,2), ..., fib(7,21), fib(8,34)
```

Complexity

- ▶ With indexing on the first argument
 - ▶ Linear complexity (each Fibonacci number only computed once)
- ▶ Without indexing on first argument
 - ▶ Quadratic complexity (Searching for suitable pairs in `mem`)
- ▶ MCT does not apply here (propagation rules)

Confluence

- ▶ Nontrivial overlaps between `mem` and each propagation rule
- ▶ First critical pair: `fib(0, M1), fib(0, M2)`

$$\begin{array}{ccc}
 & \text{fib}(0, M1), \text{fib}(0, M2) & \\
 & / \text{ mem} \qquad \qquad \backslash \text{ f0} & \\
 \text{fib}(0, M2), M1=M2 & \text{fib}(0, M1), M1=1, \text{fib}(0, M2) & \\
 | \text{ f0} & & | \text{ mem} \\
 \text{fib}(0, M2), M1=M2, M2=1 \equiv & M1=M2, M1=1, \text{fib}(0, M2) &
 \end{array}$$

Confluence

- ▶ **Second critical pair** $\text{fib}(N, M1), \text{fib}(N, M2)$ (shown split)

```

      fib(N, M1)
      | mem
    fib(N, M2), M1=M2
      | fn
fib(N, M2), M1=M2, fib(N-1, M5), fib(N-2, M6), M2 is M5+M6

      fib(N, M2)
      | fn
fib(N, M1), fib(N-1, M3), fib(N-2, M4), M1 is M3+M4, fib(N, M2)
      | mem
M1=M2, fib(N-1, M3), fib(N-2, M4), M1 is M3+M4, fib(N, M2)

```

- ▶ The two reached states are equivalent
- ▶ Overlap with rule $f1$ analogous

Other properties

- ▶ Online: trivial
- ▶ Anytime
 - ▶ In theory: no computation steps redone when started on intermediate result
 - ▶ In practice: recomputation may occur (propagation history not explicit)
 - ▶ Additional computations absorbed (confluence and `mem` rule)
 - ▶ Execution of two recursive calls in parallel possible
 - ▶ No gain: `mem` rule will absorb multiple computations

Fibonacci numbers (program variations)

- ▶ Similar reasoning, results hold for `fib` as function with given first argument
- ▶ Exception: finite bottom-up computation

```
fn @ fib_upto(Max), fib(N1,M1), fib(N2,M2)
    ==> Max>N2, N2:=N1+1 | fib(N2+1,M1+M2).
```

- ▶ Quadratic complexity (no indexing between to `fib` constraints in head)

Depth-first search

Depth-first search program

```
empty @ dfsearch(nil,X) <=> false.  
found @ dfsearch(node(N,L,R),X) <=> X=N | true.  
left @ dfsearch(node(N,L,R),X) <=> X<N | dfsearch(L,X).  
right @ dfsearch(node(N,L,R),X) <=> X>N | dfsearch(R,X).
```

- ▶ Tree encoding `node(Data, Lefttree, Righttree)`
- ▶ Data ordered such that every node in left subtree smaller, every node in right subtree larger than parent node
- ▶ Search for datum `Data` in binary tree `Tree` by calling `dfsearch(Tree, Data)`
- ▶ All analyzed properties hold in a trivial way (single-headed simplification rules with exclusive heads and guards)
- ▶ Complexity linear in depth in tree per search

Depth-first search

Depth-first search program (variant)

```
empty @ nil(I) \ dfsearch(I,X) <=> fail.  
found @ node(I,N,L,R) \ dfsearch(I,X) <=> X=N | true.  
left @ node(I,N,L,R) \ dfsearch(I,X) <=> X<N | dfsearch(L,X).  
right @ node(I,N,L,R) \ dfsearch(I,X) <=> X>N | dfsearch(R,X).
```

- ▶ Different granularity: node represented by CHR data constraint
- ▶ Tree is set of such `node` constraints
- ▶ For valid binary search tree properties of previous programs inherited
- ▶ With indexing complexity unaffected
- ▶ Data constraints can be added \Rightarrow online algorithm

Depth-first search

Depth-first search program (another variant)

```
found @ node(N) \ search(N) <=> true.  
empty @ search(N) <=> fail.
```

- ▶ Directly access data by mentioning in rule head
- ▶ All properties except anytime break down (due to `empty` rule)
- ▶ With indexing constant time complexity

Destructive assignment

Destructive assignment program

```
assign(Var,New), cell(Var,Old) <=> cell(Var,New).
```

- ▶ Constraint `assign` assigns new value to variable `Var`
- ▶ Not confluent
 - ▶ Nonjoinable overlap:
`assign(Var,New1), assign(Var,New2), cell(Var,Old)`
 - ▶ Results in either `cell(Var,New1)` or `cell(Var,New2)`
- ▶ Order matters \Rightarrow not executable in parallel as intended
- ▶ First-order logical reading does not reflect intended meaning (linear-logic semantics needed)

Transitive closure

Transitive closure program

```
dp @ p(X,Y) \ p(X,Y) <=> true.  
p1 @ e(X,Y) ==> p(X,Y).  
pn @ e(X,Y), p(Y,Z) ==> p(X,Z).
```

- ▶ Relation: edge e between two nodes
- ▶ Transitive closure: path p between two nodes

Example computation

Query $e(1,2), e(2,3), e(2,4)$ adds path constraints

$p(1,4), p(2,4), p(1,3), p(2,3), p(1,2)$

Logical reading and correctness

- ▶ First-order logical reading as implications

$$p(X, Y) \wedge p(X, Y) \leftrightarrow p(X, Y)$$

$$e(X, Y) \rightarrow p(X, Y)$$

$$e(X, Y) \wedge p(Y, Z) \rightarrow p(X, Z)$$

- ▶ Logical reading of duplicate removal is tautology
- ▶ Not expressible in FOL but in linear logic: transitive closure is smallest transitive relation
- ▶ Rules actually calculate smallest relation (left to right application produces relation bottom-up)

⇒ Program is correct

Termination

- ▶ Refined semantics
 - ▶ Duplicates removed by dp before propagation rules applied
 - ▶ Finite number of paths in finite graph

⇒ Program terminates
- ▶ Abstract semantics
 - ▶ dp can be applied too late in cyclic graph
 - ▶ Same paths generated again and again

⇒ Termination not guaranteed

Complexity (I)

- ▶ It holds that $v/2 \leq e \leq p \leq v^2$ (v #vertices, e #edges, p #paths)
- ▶ Rules can be applied in constant time
- ▶ Without indexing
 - ▶ Upper bound for propagation rule attempts: product of number of head constraints occurring during computation
 - ▶ p_1 tried at most e times, applies e times
 - ▶ p_n tried at most ep times, applies at most $\max(ev, vp) = vp$ times
 - ▶ Path constraint produced with each rule application
 - ▶ Thus, d_p applied vpv times

⇒ Worst-case complexity due to d_p $O(vp^2) = O(v^5)$

Complexity (II)

- ▶ With indexing
 - ▶ Index constraints on arguments with shared variables in heads
 - ▶ Upper bounds for rule attempts and rule application coincide now
 - ▶ p_1 tried and applied at most e times
 - ▶ p_n tried and applied at most $\max(ev, vp) = vp$ times
 - ▶ Thus, dp applied vp times now
- ⇒ Worst-case complexity due to p_n $O(vp) = O(v^3)$
- ▶ Optimal for this algorithm

Confluence

- Only nontrivial overlap (between dp and pn)

$$\begin{array}{ccc}
 e(X, Y), p(Y, Z), p(Y, Z) & & \\
 / dp & & \backslash pn \\
 e(X, Y), p(Y, Z) & e(X, Y), p(Y, Z), p(Y, Z), p(X, Z) & \\
 \backslash pn & & / dp \\
 e(X, Y), p(Y, Z), p(X, Z) & &
 \end{array}$$

- Program is confluent

Remaining properties

- ▶ Anytime: Repeated application of propagation rule does not matter
 - ▶ Confluence, duplicate paths removed
- ▶ Online: edges can be added during computation
- ▶ Strong parallelism
 - ▶ Apply p_1 to all edges in parallel
 - ▶ Next rounds: all possible applications of p_n and then d_p
 - ▶ With indexing v_p application of those rules
 - ▶ Given v processors parallel complexity $O(v^2)$
 - ▶ Cost $O(v^3)$

Single-source and single-target paths

Transitive closure program(single-source)

```
dp @ p(X,Y) \ p(X,Y) <=> true.  
sl @ source(X), e(X,Y) ==> p(X,Y).  
sn @ source(X), p(X,Y), e(Y,Z) ==> p(X,Z).
```

- ▶ Only paths from (or to) a certain node computed
- ▶ Complexity
 - ▶ Number of created path constraints reduced by factor v
($p \leq v \leq 2e \leq 2v^2$)
 - ▶ Without indexing $O(vp^2) = O(v^3)$
 - ▶ With indexing: $O(vp) = O(v^2)$

Shortest path

Shortest path program

```
dp @ p(X,Y,N) \ p(X,Y,M) <=> N=<M | true.  
e(X,Y) ==> p(X,Y,1).  
e(X,Y), p(Y,Z,N) ==> p(X,Z,N+1).
```

- ▶ Computes shortest path length between all pairs of nodes

Example computation

Query $e(X,Y), e(Y,Z), e(X,Z)$ adds path constraints

$p(X,Z,1), p(Y,Z,1), p(X,Y,1)$

Termination and complexity

- ▶ New active path constraint only removed by dp if equal or longer
- ▶ Otherwise old path removed (work repeated, at most v times)
⇒ Worst-case complexity with indexing $O(ev^2) = O(v^4)$
- ▶ Better complexity (i.e. ev) needs more clever scheduling
 - ▶ E.g. in Dijkstra's algorithm, computation always continues with shortest path found so far

Partial order constraint

Partial order program

```

duplicate    @ X leq Y \ X leq Y <=> true.
reflexivity  @ X leq X <=> true.
antisymmetry @ X leq Y , Y leq X <=> X=Y.
transitivity @ X leq Y , Y leq Z ==> X leq Z.

```

- Maintains nonstrict partial order relation $\text{leq} \leq$

Example computation

```

A leq B, C leq A, B leq C
A leq B, C leq A, B leq C, C leq B
A leq B, C leq A, B=C
A=B, B=C

```

Termination and complexity

- ▶ duplicate and transitivity analog to transitive closure (i.e. cubic)
 - ▶ reflexivity does not change complexity
 - ▶ With indexing
 - ▶ Application of antisymmetry triggers at most $O(v)$ constraints (all `leq` with X and Y)
 - ▶ In those constraints, one variable is replaced by other
 - ⇒ problem shrinks by one variable (at most $O(v)$ times)
 - ⇒ Thus, antisymmetry applied $O(v)$ times
 - ⇒ Trying and applying of antisymmetry: $O(v^2)$
- ⇒ Overall complexity $O(v^3)$

Remaining properties

- ▶ Algorithm is anytime and online (as discussed in chapter 4)
- ▶ Similar to transitive closure: `transitivity` can be applied in parallel to all pairs, then all other rules can be applied
- ▶ First-order logical reading

(duplicate) $\forall X, Y (X \leq Y \wedge X \leq Y \Leftrightarrow X \leq Y)$

(reflexivity) $\forall X (X \leq X \Leftrightarrow \text{true})$

(antisymmetry) $\forall X, Y (X \leq Y \wedge Y \leq X \Leftrightarrow X = Y)$

(transitivity) $\forall X, Y, Z (X \leq Y \wedge Y \leq Z \Rightarrow X \leq Z)$

- ▶ `duplicate` rule is tautology
- ▶ Other rules give axioms of partial order
- ▶ FOL reading suffices and shows correctness (see also chapter 3)

Cocke-Younger-Kasami

CYK algorithm

duplicate @ $p(A, I, J) \setminus p(A, I, J) \Leftrightarrow \text{true}$.

terminal @ $A \rightarrow T, e(T, I, J) \Rightarrow p(A, I, J)$.

nonterminal @ $A \rightarrow B * C, p(B, I, J), p(C, J, K) \Rightarrow p(A, I, K)$.

- ▶ Parses a string according to a context-free grammar bottom-up.
- ▶ Specialization of transitive closure

Termination and complexity

General idea: With indexing:

- ▶ Arguments of constraints can be associated with finite domains
⇒ Product of domain sizes of variables in rule head gives upper bound on number of rule applications and attempts
- ▶ Chain representing string has v nodes and $e(= v - 1)$ edges
- ▶ Grammar with t terminals and n nonterminals
- ▶ Number of grammar rules $r \leq nt + n^3$ (assuming $t \leq n^2$)
- ▶ Products of domain sizes
 - ▶ terminal (variables A, T, I, J): $ntvv = nt v^2$
 - ▶ nonterminal (variables A, B, C, I, J, K): $n^3 v^3$
 - ▶ duplicate tried with each p produced: $n^3 v^3$

⇒ Overall complexity of $O(n^3 v^3)$ with indexing (n usually fixed)

Confluence

- ▶ Confluent when used on ground chains
- ▶ Not confluent in general
- ▶ Nonjoinable critical pair from overlap

$$\begin{array}{c}
 A \rightarrow B * B, p(B, I, I), p(B, I, I) \\
 \quad / \text{ nonterminal} \qquad \qquad \quad \backslash \text{ duplicate} \\
 A \rightarrow B * B, p(B, I, I), p(B, I, I), p(A, I, I) \quad A \rightarrow B * B, p(B, I, I) \\
 \quad \quad \quad | \text{ duplicate} \\
 A \rightarrow B * B, p(B, I, I), p(A, I, I)
 \end{array}$$

Mergesort

Merge sort program

$A \rightarrow B \setminus A \rightarrow C \Leftrightarrow A < B, B = < C \mid B \rightarrow C.$

- ▶ Implements merge sort algorithm
- ▶ Query contains only arcs $0 \rightarrow A_i$
- ▶ Answer: sequence of values stored as arcs
(e.g $0, 2, 5$ is $0 \rightarrow 2, 2 \rightarrow 5$)

Example computation

$0 \rightarrow 2, 0 \rightarrow 5, 0 \rightarrow 1, 0 \rightarrow 7.$

$0 \rightarrow 2, 2 \rightarrow 5, 0 \rightarrow 1, 0 \rightarrow 7.$

$1 \rightarrow 2, 2 \rightarrow 5, 0 \rightarrow 1, 0 \rightarrow 7.$

$1 \rightarrow 2, 2 \rightarrow 5, 0 \rightarrow 1, 1 \rightarrow 7.$

$1 \rightarrow 2, 2 \rightarrow 5, 0 \rightarrow 1, 2 \rightarrow 7.$

$1 \rightarrow 2, 2 \rightarrow 5, 0 \rightarrow 1, 5 \rightarrow 7.$

Logical reading and correctness

- ▶ Classical logical reading is sufficient

$$A < B \wedge B \rightarrow C \rightarrow (A \rightarrow B \wedge A \rightarrow C \leftrightarrow A \rightarrow B \wedge B \rightarrow C).$$

- ▶ $A \rightarrow B$ means $A \leq B$, thus logical correctness is consequence of axioms for \leq

$$A < B \wedge B \leq C \rightarrow (A \leq B \wedge A \leq C \leftrightarrow A \leq B \wedge B \leq C)$$

Termination and complexity

- ▶ Complexity of merging two ordered chains (lengths n and m)
 - ▶ Indexing on the first argument of arc constraint:
 - ⇒ Second arc constraint found in constant time since rule is applicable to arbitrary pairs of arcs with same first argument
 - ▶ Each rule application processes one arc constraint
 - ⇒ $O(m + n)$
- ▶ Complexity of sorting n values given as second argument of arc
 - ▶ First argument can be replaced at most n times in each arc
 - ⇒ Worst time complexity $O(n^2)$

Confluence

- ▶ Ground confluent (correct, unique result)
- ▶ Overlaps and joinability analog to gcd
(gcd (N) mapped to $X \rightarrow N$, gcd (M-N) to $N \rightarrow M$)
- ▶ One nonjoinable overlap

$$\begin{array}{ccc}
 X \rightarrow A, X \rightarrow B, X \rightarrow C, & X < A, A = < B, & X < C, C = < B \\
 & / & | \\
 X \rightarrow A, A \rightarrow B, X \rightarrow C, & X < A, A = < B, & X < C, C = < B & | \\
 & & & | \\
 X \rightarrow A, C \rightarrow B, X \rightarrow C, & X < A, A = < B, & X < C, C = < B
 \end{array}$$

- ▶ Cannot proceed until relationship between A and C is known

Anytime and online algorithm

- ▶ Anytime property
 - ▶ Intermediate results: connected acyclic graph
 - ▶ Smallest value is root
 - ▶ Longer and longer chains without branches
- ▶ Online property
 - ▶ Sorting incrementally, new arcs can be added at any time

Mergesort (optimal complexity sorting)

- ▶ Complexity can be improved to optimal $O(n \log(n))$ by optimal merging order
- ▶ Merging chains of same length
 - ▶ Precede chain with length $(N \Rightarrow \text{Firstnode})$
 - ▶ Rule to initiate merging of chains of same length
$$N \Rightarrow A, N \Rightarrow B \Leftrightarrow A < B \mid N + N \Rightarrow A, A \rightarrow B.$$
 - ▶ Works only if length of query is a power of 2
- ▶ Start by merging n chains of length 1 then merge $n/2$ chains of length 2 and so on
- ▶ Finished after $\log(n)$ rounds \Rightarrow complexity $O(n \log(n))$
- ▶ works for any length with one more rule

Concurrency and parallelism

- ▶ Follows structure of proof of optimal complexity
- ▶ Merging of two chains strictly sequential
- ▶ In second round start merging new chains while tail of chains still produced
- ▶ $\log(n)$ rounds of merging, last round my need n more steps
- ▶ Overall $n + \log(n)$ steps
- ▶ With n processors: complexity $O(n)$ and cost $O(n^2)$
- ▶ Also possible for original version (scheduling)