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miniKanren.org/workshop/2019/8-ART1

```
> (run 12 q
    (conde
      ((repeato 'a q))
      ((repeato 'b q))
      ((repeato 'c q))
      ((repeato 'd q))))
```

$\text{cond}^e$  intuitively relates its clauses with logical or. And thus an unsuspecting beginner would expect each letter to contribute equally to the result, as follows.

```
'((a) (b) (c) (d)
  (a a) (b b) (c c) (d d)
  (a a a) (b b b) (c c c) (d d d))
```

The  $\text{cond}^e$  in TRS2, however, generates a less expected result.

```
'((a) (a a) (b) (a a a)
  (a a a a) (b b)
  (a a a a a) (c)
  (a a a a a a) (b b b)
  (a a a a a a a) (d))
```

The miniKanren in TRS2 implements interleaving DFS ( $\text{DFS}_i$ ), the cause of this unexpected result. With this search strategy, each  $\text{cond}^e$  clause takes half of its received computational resources and passes the other half to its following clauses, except for the last clause that takes all resources it receives. In the example above, the a clause takes half of all resources. And the b clause takes a quarter. Thus c and d barely contribute to the result.

$\text{DFS}_i$  is sometimes powerful for an expert. By carefully organizing the order of  $\text{cond}^e$  clauses, a miniKanren program can explore more “interesting” clauses than those uninteresting ones, and thus use computational resources efficiently.

$\text{DFS}_i$  is not always the best choice. For instance, it might be less desirable for novice miniKanren users—understanding implementation details and fiddling with clause order is not their first priority. There is another reason that miniKanren could use more search strategies than just  $\text{DFS}_i$ . In many applications, there does not exist one order that serves all purposes. For example, a relational dependent type checker contains clauses for constructors that build data and clauses for eliminators that use data. When the type checker is generating simple and shallow programs, the clauses for constructors had better be at the top of the  $\text{cond}^e$  expression. When performing proof searches for complicated programs, the clauses for eliminators had better be at the top of the  $\text{cond}^e$  expression. With  $\text{DFS}_i$ , these two uses cannot be efficient at the same time. In fact, to make one use efficient, the other one must be more sluggish. Boskin et al. [1] propose and implement a means to eliminate or re-order disjunctive clauses to accommodate varying search needs such as these.

The specification that gives every clause in the same  $\text{cond}^e$  equal “search priority” is fair  $\text{disj}$ . And search strategies with almost-fair  $\text{disj}$  give every clause similar priority. Fair  $\text{conj}$ , a related concept, is more subtle. We cover it in the next section.

Our research compares four search strategies with different features in fairness (Table 1). To summarize our contributions, we

- propose and implement **b**alanced **i**nterleaving depth-first search ( $\text{DFS}_{bi}$ )
- propose and implement **f**air depth-first search ( $\text{DFS}_f$ )
- implement in a new way the standard breath-first search. We refer to  $\text{BFS}_{ser}$  as the original implementation by Seres et al. [9] and  $\text{BFS}_{imp}$  as our new one. When we use BFS without subscripts, we mean both  $\text{BFS}_{ser}$

and  $\text{BFS}_{imp}$ . We formally prove that the two implementations are semantically equivalent, however,  $\text{BFS}_{imp}$  runs faster in all benchmarks and is shorter.

Search Strategies	disj	conj
$\text{DFS}_i$	unfair	unfair
$\text{DFS}_{bi}$	almost-fair	unfair
$\text{DFS}_f$	fair	unfair
BFS	fair	fair

Table 1. Fairness of all search strategies

## 2 SEARCH STRATEGIES AND FAIRNESS

In this section, we define fairness levels in disjunction and conjunction implementations. Specifically, a disjunction implementation can be fair, almost-fair, or unfair. And a conjunction implementation can be fair or unfair. Fairness, intuitively, measures how evenly a search strategy allocates computational resource to “sibling” spaces.

Before going further into fairness, we give a short review of the terms: *state*, *space*, and *goal*. A *state* is a collection of constraints. (Here, we restrict constraints to unification constraints.) Every answer corresponds to a state. A *space* is a collection of states. And a *goal* is a function from a state to a space. Every state in the output space includes the input state and possibly more constraints.

Now we elaborate fairness by running more queries about `repeato`. We never use `run*` here because fairness is more interesting when we ask a bounded number of answers. It is perfectly fine, however, to use `run*` with any search strategy.

### 2.1 Fair disj

Given the following program, it is natural to expect lists of each letter to constitute 1/4 in the answer list.  $\text{DFS}_i$ , TRS2’s search strategy, however, results in many more lists of `a` than lists of other letters. And some letters (e.g. `c` and `d`) are rarely seen. The more clauses, the worse the situation.

```
;; DFSi (unfair disj)
> (run 12 q
  (conde
    ((repeato 'a q))
    ((repeato 'b q))
    ((repeato 'c q))
    ((repeato 'd q))))
'((a) (a a) (b) (a a a)
  (a a a a) (b b)
  (a a a a a) (c)
  (a a a a a a) (b b b)
  (a a a a a a a) (d))
```

Under the hood, the `conde` here allocates computational resources to four trivially different spaces. The unfair `disj` in  $\text{DFS}_i$  allocates many more resources to the first space. On the contrary, fair `disj` would allocate resources evenly to each space.

<pre>;; DFS<sub>f</sub> (fair disj) &gt; (run 12 q   (cond<sup>e</sup>     ((repeat<sup>o</sup> 'a q))     ((repeat<sup>o</sup> 'b q))     ((repeat<sup>o</sup> 'c q))     ((repeat<sup>o</sup> 'd q)))) '((a) (b) (c) (d)   (a a) (b b) (c c) (d d)   (a a a) (b b b) (c c c) (d d d))</pre>	<pre>;; BFS (fair disj) &gt; (run 12 q   (cond<sup>e</sup>     ((repeat<sup>o</sup> 'a q))     ((repeat<sup>o</sup> 'b q))     ((repeat<sup>o</sup> 'c q))     ((repeat<sup>o</sup> 'd q)))) '((a) (b) (c) (d)   (a a) (b b) (c c) (d d)   (a a a) (b b b) (c c c) (d d d))</pre>
---	---

Running the same program again with almost-fair disj (e.g. DFS<sub>bi</sub>) gives the same result. Almost-fair, however, is not completely fair, as shown by the following example.

```
;; DFSbi (almost-fair disj)
> (run 16 q
  (conde
    ((repeato 'a q))
    ((repeato 'b q))
    ((repeato 'c q))
    ((repeato 'd q))
    ((repeato 'e q))))
'((b) (c) (d) (a)
  (b b) (c c) (d d) (e)
  (b b b) (c c c) (d d d) (a a)
  (b b b b) (c c c c) (d d d d) (e e))
```

DFS<sub>bi</sub> is fair only when the number of goals is a power of 2, otherwise, it allocates some goals with twice as many resources as the others. In the above example, where the cond<sup>e</sup> has five clauses, DFS<sub>bi</sub> allocates more resources to the clauses of b, c, and d.

We end this subsection with precise definitions of all levels of disj fairness. Our definition of *fair* disj is slightly more general than the one in Seres et al. [9], which is only for binary disjunction. We generalize it to a multi-arity one.

**DEFINITION 2.1 (FAIR disj).** *A disj is fair if and only if it allocates computational resources evenly to spaces produced by goals in the same disjunction (i.e., clauses in the same cond<sup>e</sup>).*

**DEFINITION 2.2 (ALMOST-FAIR disj).** *A disj is almost-fair if and only if it allocates computational resources so evenly to spaces produced by goals in the same disjunction that the maximal ratio of resources is bounded by a constant.*

**DEFINITION 2.3 (UNFAIR disj).** *A disj is unfair if and only if it is not almost-fair.*

## 2.2 Fair conj

Given the following program, it is natural to expect lists of each letter to constitute 1/4 in the answer list. Search strategies with unfair conj: DFS<sub>i</sub>, DFS<sub>bi</sub>, and DFS<sub>f</sub>, however, results in many more lists of a than lists of other letters. And some letters are rarely seen. Here again, as the number of clauses grows, the situation worsens.

Although some strategies have a different level of fairness in `disj`, they have similar behavior when there is no call to a relational definition in `conde` clauses, including this case. (The only potential difference is that `DFSbi` might reorder states.)

<pre>;; DFS<sub>i</sub> (unfair conj) &gt; (run 12 q    (fresh (x)     (cond<sup>e</sup>       ((≡ 'a x))       ((≡ 'b x))       ((≡ 'c x))       ((≡ 'd x)))     (repeat<sup>o</sup> x q))) '((a) (a a) (b) (a a a)   (a a a a) (b b)   (a a a a a) (c)   (a a a a a a) (b b b)   (a a a a a a a) (d))</pre>	<pre>;; DFS<sub>f</sub> (unfair conj) &gt; (run 12 q    (fresh (x)     (cond<sup>e</sup>       ((≡ 'a x))       ((≡ 'b x))       ((≡ 'c x))       ((≡ 'd x)))     (repeat<sup>o</sup> x q))) '((a) (a a) (b) (a a a)   (a a a a) (b b)   (a a a a a) (c)   (a a a a a a) (b b b)   (a a a a a a a) (d))</pre>	<pre>;; DFS<sub>bi</sub> (unfair conj) &gt; (run 12 q    (fresh (x)     (cond<sup>e</sup>       ((≡ 'a x))       ((≡ 'b x))       ((≡ 'c x))       ((≡ 'd x)))     (repeat<sup>o</sup> x q))) '((a) (a a) (c) (a a a)   (a a a a) (c c)   (a a a a a) (b)   (a a a a a a) (c c c)   (a a a a a a a) (d))</pre>
---	---	--

Under the hood, the `conde` and the call to `repeato` are connected by `conj`. The `conde` goal outputs a space including four trivially different states. Applying the next conjunctive goal, `(repeato x q)`, produces four trivially different spaces. In the examples above, all search strategies allocate more computational resources to the space of `a`. On the contrary, fair `conj` would allocate resources evenly to each space. For example,

```
;; BFS (fair conj)
> (run 12 q
   (fresh (x)
    (conde
      ((≡ 'a x))
      ((≡ 'b x))
      ((≡ 'c x))
      ((≡ 'd x)))
    (repeato x q)))
'((a) (b) (c) (d)
  (a a) (b b) (c c) (d d)
  (a a a) (b b b) (c c c) (d d d))
```

A more interesting situation is when the first conjunct produces an unbounded number of states. Consider the following example: a naive specification of fair `conj` might require search strategies to produce all sorts of singleton lists, but there would not be any lists of length two or longer, which makes the strategies incomplete. A search strategy is *complete* if and only if “every correct answer would be discovered after some finite time” [9], otherwise, it is *incomplete*. In the context of miniKanren, a search strategy is complete means that every correct answer has a position in large enough answer lists.

```

;; naively fair conj
> (run 6 q
  (fresh (xs)
    (conde
      ((repeato 'a xs))
      ((repeato 'b xs)))
    (repeato xs q)))
'(((a)) ((b)))
  ((a a)) ((b b))
  ((a a a)) ((b b b)))

```

Our solution requires a search strategy with *fair conj* to organize states in buckets in spaces, where each bucket is a finite collection of states and every space contains possibly infinite buckets, and to allocate resources evenly among spaces derived from states in the same bucket. It is up to a search strategy designer to decide by what criteria to put states in the same bucket, and how to allocate resources among spaces related to different buckets.

BFS puts states of the same cost in the same bucket, and allocates resources carefully among spaces related to different buckets such that it produces answers in increasing order of cost. The *cost* of an answer is its depth in the search tree (i.e., the number of calls to relational definitions required to find the answer) [9]. In the above examples, the cost of each answer is equal to their lengths because we need to apply *repeat<sup>o</sup>*  $n$  times to find an answer of length  $n$ . In the following example, every answer is a list of a list of symbols, where inner lists in the same outer list are identical. Here the cost of each answer is equal to the length of its inner lists plus the length of its outer list. For example, the cost of  $((a) (a))$  is  $1 + 2 = 3$ .

```

;; BFS (fair conj)
> (run 12 q
  (fresh (xs)
    (conde
      ((repeato 'a xs))
      ((repeato 'b xs)))
    (repeato xs q)))
'(((a)) ((b)))
  ((a) (a)) ((b) (b))
  ((a a)) ((b b))
  ((a) (a) (a)) ((b) (b) (b))
  ((a a) (a a)) ((b b) (b b))
  ((a a a)) ((b b b)))

```

We end this subsection with precise definitions of all levels of conj fairness.

**DEFINITION 2.4 (FAIR conj).** A conj is *fair* if and only if it allocates computational resources evenly to spaces produced from states in the same bucket. A bucket is a finite collection of states. And search strategies with *fair conj* should represent spaces with possibly unbounded collections of buckets.

**DEFINITION 2.5 (UNFAIR conj).** A conj is *unfair* if and only if it is not fair.

```

#| Goal × Goal → Goal |#
(define (disj2 g1 g2)
  (lambda (s)
    (append∞ (g1 s) (g2 s))))

#| Space × Space → Space |#
(define (append∞ s∞ t∞)
  (cond
    ((null? s∞) t∞)
    ((pair? s∞)
     (cons (car s∞)
           (append∞ (cdr s∞) t∞)))
    (else (lambda ()
              (append∞ t∞ (s∞))))))

(define-syntax disj
  (syntax-rules ()
    ((disj) (fail))
    ((disj g0 g ...) (disj+ g0 g ...))))

(define-syntax disj+
  (syntax-rules ()
    ((disj+ g) g)
    ((disj+ g0 g1 g ...) (disj2 g0 (disj+ g1 g ...))))

```

Fig. 1. implementation of DFS<sub>i</sub> (Part I)

### 3 INTERLEAVING DEPTH-FIRST SEARCH

In this section, we review the implementation of interleaving depth-first search (DFS<sub>i</sub>). We focus on parts that are relevant to other strategies. TRS2, chapter 10 and the appendix, “Connecting the wires”, provide a comprehensive description of the miniKanren implementation but limited to unification constraints ( $\equiv$ ). Fig. 1 and Fig. 2 show parts that are later compared with other search strategies. We follow some conventions to name variables: *ss* name states; *gs* (possibly with subscript) name goals; variables ending with <sup>∞</sup> name spaces. Fig. 1 shows the implementation of *disj*. The first function, *disj<sub>2</sub>*, implements binary disjunction. It applies the two disjunctive goals to the input state *s* and composes the two resulting spaces with *append<sup>∞</sup>*. The following syntax definitions say *disj* is right-associative. Fig. 2 shows the implementation of *conj*. The first function, *conj<sub>2</sub>*, implements binary conjunction. It applies the *first* goal to the input state, then applies the second goal to states in the resulting space. The helper function *append-map<sup>∞</sup>* applies its input goal to states in its input space and composes the resulting spaces. It reuses *append<sup>∞</sup>* for space composition. The following syntax definitions say *conj* is also right-associative.

```

#| Goal × Goal → Goal |#
(define (conj2 g1 g2)
  (lambda (s)
    (append-map∞ g2 (g1 s))))

#| Goal × Space → Space |#
(define (append-map∞ g s∞)
  (cond
    ((null? s∞) '())
    ((pair? s∞)
     (append∞ (g (car s∞))
               (append-map∞ g (cdr s∞))))
    (else (lambda ()
              (append-map∞ g (s∞))))))

(define-syntax conj
  (syntax-rules ()
    ((conj) (fail))
    ((conj g0 g ...) (conj+ g0 g ...))))

(define-syntax conj+
  (syntax-rules ()
    ((conj+ g) g)
    ((conj+ g0 g1 g ...) (conj2 g0 (conj+ g1 g ...)))))

```

Fig. 2. implementation of DFS<sub>i</sub> (Part II)

#### 4 BALANCED INTERLEAVING DEPTH-FIRST SEARCH

The strategy **balanced interleaving** DFS (DFS<sub>bi</sub>) has an almost-fair `disj` and unfair `conj`. The implementation of DFS<sub>bi</sub> differs from DFS<sub>i</sub>'s in the `disj` macro. When there are one or more disjunctive goals, the new `disj` builds a balanced binary tree whose leaves are the goals and whose nodes are `disj2s`, hence the name of this search strategy. In contrast, the `disj` in DFS<sub>i</sub> constructs the binary tree in a particularly unbalanced form. We list the new `disj` with its helper in Fig. 3. The new helper, `disj+`, takes two additional 'arguments'. They accumulate goals to be put in the left and right subtrees. The first clause handles the case where there is only one goal. In this case, the tree is the goal itself. When there are more goals, we partition the list of goals into two sublists of roughly equal lengths and recur on the two sublists. We move goals to the accumulators in the last clause. As we are moving two goals each time, there are two base cases: (1) no goal remains; (2) one goal remains. We handle these two new base cases in the second clause and the third clause, respectively.

#### 5 FAIR DEPTH-FIRST SEARCH

Fair DFS (DFS<sub>f</sub>) has fair `disj` and unfair `conj`. The implementation of DFS<sub>f</sub> differs from DFS<sub>i</sub>'s in `disj2` (Fig. 4). The new `disj2` calls a new and fair version of `append∞`. `append∞fair` immediately calls its helper, `loop`, with the



```

(define-syntax disj
  (syntax-rules ()
    ((disj) fail)
    ((disj g ...) (disj+ (g ...) () ())))))

(define-syntax disj+
  (syntax-rules ()
    ((disj+ () () g) g)
    ((disj+ (gl ...) (gr ...))
     (disj2 (disj+ () () gl ...)
              (disj+ () () gr ...))))
    ((disj+ (gl ...) (gr ...) g0)
     (disj2 (disj+ () () gl ... g0)
              (disj+ () () gr ...))))
    ((disj+ (gl ...) (gr ...) g0 g1 g ...)
     (disj+ (gl ... g0) (gr ... g1) g ...))))

```

Fig. 3. implementation of  $\text{DFS}_{bi}$ 

```

#| Goal  $\times$  Goal  $\rightarrow$  Goal |#
(define (disj2 g1 g2)
  (lambda (s)
    (appendfair $\infty$  (g1 s) (g2 s))))

#| Space  $\times$  Space  $\rightarrow$  Space |#
(define (appendfair $\infty$  s $\infty$  t $\infty$ )
  (let loop ((s? #t) (s $\infty$  s $\infty$ ) (t $\infty$  t $\infty$ ))
    (cond
      ((null? s $\infty$ ) t $\infty$ )
      ((pair? s $\infty$ )
       (cons (car s $\infty$ )
              (loop s? (cdr s $\infty$ ) t $\infty$ )))
      (s? (loop #f t $\infty$  s $\infty$ ))
      (else (lambda ()
                (loop #t (t $\infty$ ) (s $\infty$ )))))))

```

Fig. 4. implementation of  $\text{DFS}_f$ 

first argument,  $s?$ , initialized to  $\#t$ , which indicates that we haven't swapped  $s^\infty$  and  $t^\infty$ . The swapping happens at the third `cond` clause in `loop`, where  $s?$  is updated accordingly. The first two `cond` clauses essentially copy the cars and stop recursion when one of the input spaces is obviously finite. The third clause, as we mentioned

```

#| Goal × Space → Space |#
(define (append-map∞fair g s∞)
  (cond
    ((null? s∞) '())
    ((pair? s∞)
     (append∞fair (g (car s∞))
                   (append-map∞fair g (cdr s∞))))
    (else (lambda ()
              (append-map∞fair g (s∞))))))

```

Fig. 5. stepping-stone toward  $\text{BFS}_{imp}$  (based on  $\text{DFS}_f$ )

above, is just for swapping. When the fourth and last clause runs, we know that both  $s^\infty$  and  $t^\infty$  end with thunks, and that we have swapped them. In this case, we construct a new thunk. The new thunk swaps back the two spaces in the recursive call to loop. This is unnecessary for fairness—we do it to produce answers in a more readable order.

## 6 BREADTH-FIRST SEARCH

BFS has both fair  $\text{disj}$  and fair  $\text{conj}$ . Our first BFS implementation (Fig. 5) serves as a “stepping-stone” toward  $\text{BFS}_{imp}$ . It is so similar to  $\text{DFS}_f$  (not  $\text{DFS}_i$ ) that we only need to apply two changes: (1) rename  $\text{append-map}^\infty$  to  $\text{append-map}^\infty_{fair}$  and (2) replace  $\text{append}^\infty$  with  $\text{append}^\infty_{fair}$  in  $\text{append-map}^\infty_{fair}$ ’s body.

This implementation can be improved in two ways. First, as mentioned in subsection 2.2, BFS puts answers in buckets and answers of the same cost are in the same bucket. In the above implementation, however, it is not obvious how we manage cost information—the cars of a space have cost 0 (i.e., they are all in the same bucket), and every thunk indicates an increment in cost. It is even more subtle that  $\text{append}^\infty_{fair}$  and the  $\text{append-map}^\infty_{fair}$  respects the cost information. Second,  $\text{append}^\infty_{fair}$  is extravagant in memory usage. It makes  $O(n + m)$  new cons cells every time, where  $n$  and  $m$  are the sizes of the first buckets of two input spaces.  $\text{DFS}_f$  is also space extravagant.

In the following paragraphs, we first describe  $\text{BFS}_{imp}$  implementation that manages cost information in a more clear and concise way and is less extravagant in memory usage. Then we compare  $\text{BFS}_{imp}$  with  $\text{BFS}_{ser}$ .

We simplify the cost information by changing the Space type, modifying related function definitions, and introducing a few more functions. The new type of Space is a pair whose car is a list of answers (the bucket), and whose cdr is either #f or a thunk returning a space. The #f here means the space is obviously finite, just like empty list in other implementations. We list functions related to the pure subset in Fig. 6. The first three functions are space constructors. none makes an empty space; unit makes a space from one answer; and step makes a space from a thunk. The remaining functions are as before. Luckily, the change in  $\text{append}^\infty_{fair}$  also fixes the miserable space extravagance—the use of append helps us to reuse the first bucket of  $t^\infty$ . Functions implementing impure features are in Fig. 7. The first function, elim, takes a space  $s^\infty$  and two continuations  $fk$  and  $sk$ . When  $s^\infty$  contains no answers, it calls  $fk$ . Otherwise, it calls  $sk$  with the first answer and the rest of the space. This function is similar to an eliminator of spaces, hence the name. The remaining functions are as before.

Kiselyov et al. [6] have demonstrated that a *MonadPlus* hides in implementations of logic programming systems.  $\text{BFS}_{imp}$  is not an exception:  $\text{append-map}^\infty_{fair}$  is like bind, but takes arguments in reversed order; none, unit, and  $\text{append}^\infty_{fair}$  correspond to mzero, unit, and mplus, respectively.

```

#|  $\rightarrow \text{Space}$  |#
(define (none)
  `(() . #f))

#|  $\text{State} \rightarrow \text{Space}$  |#
(define (unit s)
  `((,s) . #f))

#|  $(\rightarrow \text{Space}) \rightarrow \text{Space}$  |#
(define (step f)
  `(() . ,f))

#|  $\text{Space} \times \text{Space} \rightarrow \text{Space}$  |#
(define (appendfair∞ s∞ t∞)
  (cons (append (car s∞) (car t∞))
    (let ((t1 (cdr s∞)) (t2 (cdr t∞)))
      (cond
        ((not t1) t2)
        ((not t2) t1)
        (else (lambda () (appendfair∞ (t1) (t2))))))))))

#|  $\text{Goal} \times \text{Space} \rightarrow \text{Space}$  |#
(define (append-mapfair∞ g s∞)
  (foldr
    (lambda (s t∞)
      (appendfair∞ (g s) t∞))
    (let ((f (cdr s∞)))
      (step (and f (lambda () (append-mapfair∞ g (f))))))
    (car s∞)))

#|  $\text{Maybe Nat} \times \text{Space} \rightarrow [\text{State}]$  |#
(define (take∞ n s∞)
  (let loop ((n n) (vs (car s∞)))
    (cond
      ((and n (zero? n)) '())
      ((pair? vs)
       (cons (car vs)
         (loop (and n (sub1 n)) (cdr vs)))))
    (else (let ((f (cdr s∞)))
      (if f (take∞ n (f)) '()))))))))

```

Fig. 6. New and changed functions in  $\text{BFS}_{\text{imp}}$  that implements pure features

```

#| Space × (State × Space → Space) × (→ Space) → Space |#
(define (elim s∞ fk sk)
  (let ((ss (car s∞)) (f (cdr s∞)))
    (cond
      ((pair? ss) (sk (car ss) (cons (cdr ss) f)))
      (f (step (lambda () (elim (f) fk sk))))
      (else (fk)))))

#| Goal × Goal × Goal → Goal |#
(define (ifte g1 g2 g3)
  (lambda (s)
    (elim (g1 s)
      (lambda () (g3 s))
      (lambda (s0 s∞)
        (append-mapfair∞ g2
          (appendfair∞ (unit s0) s∞))))))

#| Goal → Goal |#
(define (once g)
  (lambda (s)
    (elim (g s)
      (lambda () (none))
      (lambda (s0 s∞) (unit s0)))))

```

Fig. 7. New and changed functions in  $BFS_{imp}$  that implement impure features

Now we compare the pure subset of  $BFS_{imp}$  with  $BFS_{ser}$ . We focus on the pure subset because  $BFS_{ser}$  is designed for a pure relational programming system. We prove in Coq that these two search strategies are semantically equivalent, since the result of  $(run\ n\ ?\ g)$  is the same either way. (See supplements for the formal proof.) To compare efficiency, we translate  $BFS_{ser}$ 's Haskell code into Racket (See supplements for the translated code). The translation is direct due to the similarity of the two relational programming systems. The translated code is longer than  $BFS_{imp}$ . And it runs slower in all benchmarks. Details about differences in efficiency are in section 7.

## 7 QUANTITATIVE EVALUATION

In this section, we compare the efficiency of the search strategies. A concise description is in Table 2. A hyphen means “running out of 500 MB memory”. The first two benchmarks are from TRS2. `reverso` is from Rozplokh and Boulytchev [8]. The next two benchmarks about generating quines are based on a similar test case in Byrd et al. [2]. We modify the relational interpreters because we don't have disequality constraints (e.g. `absento`). The sibling benchmarks differ in the `conde` clause order of their relational interpreters. The last two benchmarks are about synthesizing expressions that evaluate to `'(I love you)`. They are also based on a similar test case in Byrd et al. [2]. Again, we modify the relational interpreters for the same reason. And the sibling benchmarks differ in the `conde` clause order of their relational interpreters. The first one has elimination rules (i.e. application,

benchmark	size	DFS <sub>i</sub>	DFS <sub>bi</sub>	DFS <sub>f</sub>	BFS <sub>imp</sub>	BFS <sub>ser</sub>
very-recursive <sup>o</sup>	100000	166	103	412	451	1024
	200000	283	146	765	839	1875
	300000	429	346	2085	1809	4408
append <sup>o</sup>	100	17	18	17	17	65
	200	145	137	137	142	121
	300	388	384	387	429	371
revers <sup>o</sup>	10	3	4	3	18	36
	20	35	35	33	3070	3695
	30	329	333	315	75079	107531
quine-1	1	13	14	10	-	-
	2	30	34	25	-	-
	3	41	48	33	-	-
quine-2	1	22	21	12	-	-
	2	51	46	24	-	-
	3	72	76	32	-	-
'(I love you)-1	999	78	70	59	254	605
	1998	152	154	109	315	635
	2997	430	440	266	317	643
'(I love you)-2	999	631	173	61	253	605
	1998	1233	511	111	308	659
	2997	2145	682	269	307	642

Table 2. The results of a quantitative evaluation: running times of benchmarks in milliseconds

car, and cdr) at the end, while the other has them at the beginning. We conjecture that DFS<sub>i</sub> would perform badly in the second case because elimination rules complicate the problem when synthesizing (i.e., our evaluation supports our conjecture.)

In general, variants of DFS usually performs better than BFS. The reason might be that BFS tends to remember more states at the same time. Among the three variants of DFS, which all have unfair conj, DFS<sub>f</sub> is most resistant to clause permutation in quines and '(I love you)s, followed by DFS<sub>bi</sub> then DFS<sub>i</sub>. Thus, we consider DFS<sub>bi</sub> and DFS<sub>f</sub> competitive alternatives to DFS<sub>i</sub>. Among the two implementations of BFS, BFS<sub>imp</sub> constantly performs as well or better.

## 8 RELATED RESEARCH

In this section, we describe related research. Yang [10] points out that a disjunct complex would be ‘fair’ if it were a full and balanced tree. Seres et al. [9] describe a breadth-first search strategy. We present another implementation. Our implementation is semantically equivalent to theirs. But, ours is shorter and performs better in comparison with a straightforward translation of their Haskell code. Rozplochas and Boulytchev [8] address the non-commutativity of conjunction, while our work about disj fairness addresses the non-commutativity of disjunction.

## 9 CONCLUSION

We analyze the definitions of fair disj and fair conj, then propose a new definition of fair conj. Our definition of fair conj is orthogonal with completeness.

We devise two new search strategies (i.e., balanced interleaving DFS ( $\text{DFS}_{bi}$ ) and fair DFS ( $\text{DFS}_f$ )) and devise a new implementation of BFS ( $\text{BFS}_{imp}$ ). These strategies have different features in fairness:  $\text{DFS}_{bi}$  has an almost-fair disj and unfair conj.  $\text{DFS}_f$  has fair disj and unfair conj. BFS has both fair disj and fair conj. No search strategy here combines unfair disj and fair conj. This is because we haven't seen a case where this kind of search strategies would be interesting.

Our quantitative evaluation shows that  $\text{DFS}_{bi}$  is a competitive alternative to  $\text{DFS}_i$ , the current miniKanren search strategy, and that  $\text{BFS}_{imp}$  is more practical than  $\text{BFS}_{ser}$ .

We prove formally that  $\text{BFS}_{imp}$  is semantically equivalent to  $\text{BFS}_{ser}$ . But,  $\text{BFS}_{imp}$  is shorter and performs better in comparison with a straightforward translation of their Haskell code.

Although there are very few benchmarks, this is preliminary work where we are making a point that certain levels of fairness come without cost in some cases, and that each of the search strategies:  $\text{DFS}_i$ ,  $\text{DFS}_{bi}$ ,  $\text{DFS}_f$ , and BFS, can co-exist inside one's head. Constructing a miniKanren with all levels of fairness is future work.

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