# miniKanren with fair search strategies

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#### **ACM Reference Format:**

#### 1 INTRODUCTION

miniKanren programs, especially relational interpreters, have been proven to be useful in solving many problems [1]. A subtlety in writing relational programs having large cond<sup>e</sup> expressions, such as interpreters, is that the order of cond<sup>e</sup> clauses can affect the speed considerably. When a cond<sup>e</sup> expression is large enough, the left clauses consume almost all the resource and the right ones are hardly explored. The unfair disj of the current search strategy, interleaving DFS, is the cause. Under the hood, cond<sup>e</sup> uses conj to create a goal for each clause, and disj to combine these goals to one. The current disj allocates half resource to its first goal, then allocates the other half to the rest similarly, except for the last clause which takes all the resource.

Being aware of disj fairness, we also investigate conj fairness.

We propose three new search strategies, balanced interleaving DFS (biDFS), fair DFS (fDFS), and BFS. They have different characteristics of fairness (Table. 1). We prove that our BFS and the BFS proposed by Seres et al [3] produce the same result when queried. But our code is shorter and runs faster. We also compare the efficiency of these new search strategies with the existing ones.

#### 2 FAIRNESS

We demonstrate the aspects (disj or conj) and levels (unfair, almost-fair, or fair) of fairness by running queries about repeato, a relational definition that relates a term x with a non-empty list whose elements are x (Fig. 1).

## 2.1 fair disj

In the following program, the three cond<sup>e</sup> clauses differ in a trivial way. So we expect lists of each sort constitute 1/4 of the answer list. However, iDFS, the current search strategy, gives us much more lists of a than other sorts of lists. And some sorts (e.g. lists of c) are hardly found. The situation would be worse if we add more cond<sup>e</sup> clauses.

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fairness	iDFS	biDFS	fDFS	BFS
disj	unfair	almost-fair	fair	fair
conj	unfair	unfair	unfair	fair

Table 1. fairness of search strategies

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Fig. 1. repeato and an example run

On the contrary, search strategies with fair disj (e.g. fDFS and BFS) give a nice answer list.

Search strategies with almost-fair disj (e.g. biDFS) give the same result in this case. However, as its name suggests, almost-fair strategies are not always fair. Fortunately, the maximal ratio of the allocated resource is bounded by a constant. Our biDFS is fair when the number of goals is a power of 2, otherwise, some goals are allocated twice more resource than the others. In following example, the clauses of b, c, and d are allocated more resource.

## 2.2 fair conj

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In the following program, the three cond<sup>e</sup> clauses differ in a trivial way. So we expect lists of each sort constitute 1/4 of the answer list. However, search strategies with unfair conj gives us much more lists of a than other sorts of lists. And some sorts (e.g. lists of c) are hardly found. The situation would be worse if we add more cond<sup>e</sup> clauses. The result with biDFS, whose conj is also unfair, is similar, but due to its different disj, the position of b and c and swapped.

On the contrary, search strategies with fair conj (e.g. BFS) give a nice answer list.

```
142
                            ;; BFS (fair conj)
143
                           > (run 12 q
144
                                (fresh (x)
145
                                   (conde
146
                                     [(== 'a x)]
147
                                     [(== 'b x)]
148
                                     [(== 'c x)]
149
                                     [(== 'd x)])
150
                                   (repeato x q)))
151
                            '((a) (b) (c) (d)
152
153
                              (a a) (b b) (c c) (d d)
                              (a a a) (b b b) (c c c) (d d d))
154
155
```

A more interesting situation is when the first conjunctive goal produces infinite many states. Consider the following example, a naive specification of fair conj might require search strategies to produce all sorts of singleton lists, but no longer lists, which makes the strategies *incomplete*.

Fairness is good but is not good enough to kick away completeness. A solution also taken in [3] is bagging states and requiring that search spaces derived from states in the same bag are treated fairly. A natural way to bag states is by their costs. The *cost* of a state is its depth in the search tree (i.e. the number of calls to relational definitions required to find them) [3]. BFS is more than fair – it produces answers in increasing order of cost. Running the same program gives an answer list sorted by answers' costs. In this case, the cost of an answer is equal to the length of the inner lists plus the length of the outer list.

```
189
                              ;; BFS (fair conj)
190
                              > (run 12 q
191
                                   (fresh (xs)
192
                                     (conde
193
                                        [(repeato 'a xs)]
194
                                        [(repeato 'b xs)])
195
                                     (repeato xs q)))
196
                              '(((a)) ((b))
197
                                ((a) (a)) ((b) (b))
198
                                ((a a)) ((b b))
199
200
                                ((a) (a) (a)) ((b) (b) (b))
201
                                ((a a) (a a)) ((b b) (b b))
202
                                ((a a a)) ((b b b)))
203
```

#### 3 BALANCED INTERLEAVING DFS

Balanced interleaving DFS (biDFS) has almost-fair disj and unfair conj. The implementation of biDFS differs from iDFS in the disj macro. We list the new disj with its helpers in Fig. 2. The first helper function, split, takes a list of goals ls and a procedure k, partitions ls into two sub-lists of roughly equal length, and returns the application of k to the two sub-lists. disj\* takes a non-empty list of goals gs and returns a goal. With the help of split, it essentially constructs a balanced binary tree where leaves are goals in gs and nodes are disj2, whence the name of this search strategy. In contrast, the disj in iDFS essentially constructs the same sort of binary tree in one of the most unbalanced forms.

#### 4 FAIR DFS

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Fair DFS (fDFS) has fair disj and unfair conj. The implementation of fDFS differs from iDFS's in disj2 (Fig. 3). disj2 is changed to call a new and fair version of append-inf. append-inf/fair immediately calls its helper, append-inf/fair^, with the first argument, s?, set to #t, which indicates that s-inf and t-inf haven't been swapped. The swapping happens in the third cond clause of the helper, where s? is changed 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 early, is just for swapping. When the fourth and last clause runs, we know that both s-inf and t-inf are ended with a thunk. In this case, a new thunk is constructed. The new thunk calls the driver recursively. Here changing the order of t-inf and s-inf won't hurt the fairness (though it will change the order of answers). We swapped them back so that answers are produced in a more natural order.

## 5 BREADTH-FIRST SEARCH

Our BFS is fair in both disj and conj. Its implementation is based on fDFS (not iDFS). All we have to do is changing the use of append-inf in append-map-inf to append-inf/fair. After making this change, append-inf is useless.

The implementation can be improved in two aspects. First, as we mentioned in section 2.2, our BFS bag states by their cost. However, in this implementation, it is unclear where this information is recorded. Second, append-inf/fair is space inefficient. It makes O(n+m) new cons cells every time, where n and m is the "length" of each input search space. We address these issues in the first subsection.

```
236
     #| [Goal] x ([Goal] x [Goal] -> Goal) -> Goal |#
237
     (define (split ls k)
238
       (cond
239
          [(null? ls) (k '() '())]
240
          [else (split (cdr ls)
241
                   (lambda (l1 l2)
242
                     (k (cons (car ls) 12) 11)))]))
243
244
     #| [Goal] -> Goal |#
245
     (define (disj* gs)
246
247
       (cond
248
          [(null? (cdr gs)) (car gs)]
249
         [else
250
           (split gs
251
             (lambda (gs1 gs2)
252
               (disj2 (disj* gs1)
253
                        (disj* gs2))))]))
254
255
     (define-syntax disj
256
       (syntax-rules ()
257
          [(disj) fail]
258
          [(disj g ...) (disj* (list g
259
260
261
                                      Fig. 2. balanced-disj
```

Both our BFS and Seres's BFS [3] produce answers in increasing order of cost. So it is interesting to see if they are equivalent. We prove the equivalence in Coq. The details are in the second subsection.

#### 5.1 optimized BFS

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As we mentioned in section 2.2, our BFS bag states by their cost. The bagging information is recorded subtly – the cars of a search space have cost 0 (they are in the same bag), and the costs of states in thunk are increased by one. It is even more difficult to see append-inf/fair and the modified append-map-inf respects the cost information. We make things clear by changing the type of search space, modify related function definitions, and introducing a few more functions.

The new type is a pair whose car is a list of state (the bag), and whose cdr is either a #f or a thunk returning a search space. A falsy cdr indicates that the search space is obviously finite.

Both modified functions and introduced ones are listed in Fig. 4. The first three functions in Fig. 4 are search space constructor. They are newly introduced. none makes an empty search space. unit makes an space from one state. step makes a space from a thunk. The next function, split, is used to implement impure features (i.e. ifte and once). It takes a space s-inf and two continuations ks and kf. When s-inf contains some states, ks is called with the first state and the rest space. Otherwise, kf is called with no argument. Here 's' and 'f' means 'succeed' and 'fail' respectively. The remaining functions do the

```
283
    #| Goal x Goal -> Goal |#
284
     (define (disj2 g1 g2)
285
       (lambda (s)
286
         (append-inf/fair (g1 s) (g2 s))))
287
288
     #| Space x Space -> Space |#
289
     (define (append-inf/fair s-inf t-inf)
290
       (append-inf/fair^ #t s-inf t-inf))
291
292
     #| Bool x Space x Space -> Space |#
293
294
     (define (append-inf/fair s? s-inf t-inf)
295
       (cond
296
         ((pair? s-inf)
297
          (cons (car s-inf)
298
            (append-inf/fair^ s? (cdr s-inf) t-inf)))
299
         ((null? s-inf) t-inf)
300
         (s? (append-inf/fair^ #f t-inf s-inf))
301
         (else (lambda ()
302
                  (append-inf/fair (t-inf) (s-inf)))))
303
304
```

Fig. 3. How fDFS differs from iDFS

same thing as before. Now it should be not difficult to see that append-inf/fair and append-map-inf do respect cost information.

Luckily, the change in append-inf/fair also fixes the miserable space inefficiency – the use of append helps us to reuse the first bag of t-inf.

Noted that some functions in the list constitute a MonadPlus: none, unit, append-map-inf, and append-inf correspond to mzero, unit, bind, and mplus respectively.

## 5.2 comparison with Silvija's BFS

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328 329 In this section, we compare the pure subset of our optimized BFS with the BFS found in [3]. We focus on the pure subset because Silvija's system is pure.

To compare efficiency, we translate her Haskell code into Racket (See supplements for the translated code), and embed it into miniKanren. The translation is fairly straightforward due to the similarity in both logic programming system and search space representation. The translated code is less efficient when running our benchmark. Details about efficiency difference are in section 6.

We prove the two search strategies are equivalent in Coq. Since search space can be infinite, we should use a co-inductive data type. However, Coq is too strict in the guardedness condition to accept a direct translation of the implementations. Therefore, we prove core theorems with finite search space instead. In order to generalize the conclusion to the cases with infinite search space, we prove a few more theorems saying that whenever we query answers lower than some finite cost, we can restrict goals to truncate search spaces at some finite depth without changing the query result. (See supplements for the formal proof)

```
330
                                . #f))
     (define (none)
                        '(()
331
     (define (unit s) '((,s) . #f))
332
     (define (step f) '(()
                                . ,f))
333
334
     (define (split s-inf ks kf)
335
       (let ([ss (car s-inf)]
336
              [f (cdr s-inf)])
337
         (cond
338
           [(and (null? ss) f)
339
            (step (lambda () (split (f) ks kf)))]
340
341
           [(null? ss) (kf)]
342
           [else (ks (car ss) (cons (cdr ss) f))])))
343
344
     (define (append-inf/fair s-inf t-inf)
345
       (cons (append (car s-inf) (car t-inf))
346
         (let ([t1 (cdr s-inf)]
347
                [t2 (cdr t-inf)])
348
           (cond
349
              [(not t1) t2]
350
              [(not t2) t1]
351
              [else (lambda () (append-inf/fair (t1) (t2)))]))))
352
353
     (define (append-map-inf g s-inf)
354
355
       (foldr
356
         (lambda (s t-inf)
357
            (append-inf/fair (g s) t-inf))
358
         (let ([f (cdr s-inf)])
359
            (step (and f (lambda () (append-map-inf g (f))))))
360
         (car s-inf)))
361
362
     (define (take-inf n s-inf)
363
       (let loop ([n n]
364
                   [vs (car s-inf)])
365
         (cond
366
367
           ((and n (zero? n)) '())
368
           ((pair? vs)
369
            (cons (car vs)
370
               (loop (and n (sub1 n)) (cdr vs))))
371
           (else
372
            (let ([f (cdr s-inf)])
373
               (if f (take-inf n (f)) '())))))
374
375
376
```

Fig. 4. interface functions in optimized BFS

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benchmark	size	iDFS	biDFS	fDFS	optimized BFS	Silvija's BFS
very-recursiveo	100000	579	793	2131	1438	3617
	200000	1283	1610	3602	2803	4212
	300000	2160	2836	-	6137	-
appendo	100	31	41	42	31	68
	200	224	222	221	226	218
	300	617	634	593	631	622
reverseo	10	5	3	3	38	85
	20	107	98	51	4862	5844
	30	446	442	485	123288	132159
quine-1	1	71	44	69	-	-
	2	127	142	95	_	-
	3	114	114	93	-	-
quine-2	1	147	112	56	-	-
	2	161	123	101	_	-
	3	289	189	104	$\langle \overline{\chi}_{\overline{\chi}} \rangle$ .	-
'(I love you)-1	99	56	15	22	74	165
	198	53	72	55	47	74
	297	72	90	44	181	365
'(I love you)-2	99	242	61	16	66	99
	198	445	110	60	42	64
	297	476	146	49	186	322

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

;;TODO to be more specific?

#### 6 QUANTITATIVE EVALUATION

In this section, we compare the efficiency of search strategies. A concise description is in Table 2. A hyphen means running out of memory. The first three benchmarks are taken from [2]. Next two benchmarks about quine are modified from a similar test case in [1]. The modifications are made to circumvent the need for symbolic constraint. Our version generates de Bruijnized expressions and forbids closures going into list. The two benchmarks differ in the cond<sup>e</sup> clause order of their relation interpreters. The last two benchmarks are about synthesizing expressions that evaluate to '(I love you). This benchmark is also inspired by [1]. Again, they differ in the cond<sup>e</sup> clause order of their relation interpreters. The first one has elimination rules (i.e. application, car, and cdr) at the end, while the other has them at the beginning. We conjecture that iDFS would perform badly in the second case because elimination rules complicate the problem when running backward. Our statistics support our conjecture.

In general, only iDFS and biDFS constantly perform well. Among them, biDFS seems to be less sensitive to change in cond<sup>e</sup> clause order (see the last four benchmarks). The other search strategies all have fair disj. And they all perform badly in the very-recursiveo benchmark. However, the drawback of having fair disj alone (i.e. fDFS) is not shown elsewhere. Fair conj impose overhead constantly except in appendo. The reason might be that strategies with fair conj tend to keep more intermediate states in the memory. Among the BFSs, our version performs better in most cases, and equally well elsewhere.

#### 7 RELATED WORKS

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Edward points out a disjunct complex would be 'fair' if it is a full and balanced tree [4].

Silvija et al [3] also describe a breadth-first search strategy. We proof their BFS is equivalent to ours. However, ours looks simpler and performs better in comparison with a straightforward translation of their Haskell code.

# 8 CONCLUSION

## **ACKNOWLEDGMENTS**

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