Generalised Selection Monad

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Abstract. General setup and introduction words here

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[] {-# LANGUAGE Impredicative Types #-} {-# LANGUAGE Scoped Type-Variables #-}

[] import Prelude hiding ((>>=), return, pure, (<*>), fmap, sequence, Left, Right)

1 Introduction to the Selection Monad

This section introduces the selection monad, focusing on the type J r a = (a -> r) -> a for selection functions. The pair function is explored, showcasing its capability to compute new selection functions based on criteria from two existing functions. Illustrated with a practical example, the decision-making scenarios involving individuals navigating paths underscore the functionality of selection functions.

An analysis of the inefficiency in the original pair function identifies redundant computational work. The primary contribution of the paper is then outlined: an illustration and proposal for an efficient solution to enhance pair function performance. This introductory overview sets the stage for a detailed exploration of the selection monad and subsequent discussions on optimizations.

1.1 Selection functions

Consider the tollowing already know type for selection functions:

[] type J r
$$a = (a -> r) -> a$$

When given two selection functions, a pair function can be defined to compute a new selection function. This resultant function selects a pair based on the criteria established by the two given selection functions:

```
[] pair :: J r a -> J r b -> J r (a,b) pair f g p = (a,b) where a = f(x -> p(x, g(y -> p(x,y)))) b = g(y -> p(a,y))
```

1.2 Example to illustrate the pair function

To gain a deeper understanding of the provided pair function, consider the following example. Picture two individuals walking on a path, one heading north and the other south. As they proceed, a collision is imminent. At this juncture, each individual must make a decision regarding their next move. This decision-making process can be modeled using selection functions. The decision they need to make is modeled as either going right or left:

```
[] data Decision = Left — Right
```

The respective selection functions decide given a predicate that tells them what decision is the correct one, select the correct one, and if there is no correct one, they default to walking right.

[] p1, p2 :: J Bool Decision p1 p = if p Left then Left else Right p2 p = if p Left then Left else Right

To apply the pair function, a predicate pred is needed that will judge two decisions and return True if a crash would be avoided and False otherwise.

```
[] pred :: (Decision, Decision) -> Bool pred (Left,Right) = True pred (Right,Left) = True pred \_= False
```

With the pair function, the merging of the two selection functions into a new one that identifies an optimal decision can now be calculated.

```
pair p1 p2 pred
--> (Left,Right)
```

Examining how the pair function is defined reveals that the first element a of the pair is determined by applying the initial selection function f to a newly constructed property function. Intuitively, selection functions can be conceptualized as entities containing a collection of objects, waiting for a property function to assess their underlying elements. Once equipped with a property function, they can apply it to their elements and select an optimal one.

Considering the types assigned to selection functions, it is evident that an initial selection function f remains in anticipation of a property function of type (a \rightarrow r) to determine an optimal a. The pair function is endowed with a property function p:: ((a,b) \rightarrow r). Through the utilization of this property function, a property function for f can be derived by using the second selection function g to select a corresponding b and subsequently applying p to assess (a,b) pairs as follows: (\x \rightarrow p (x, g (\y \rightarrow p (x,y))). Upon the determination of an optimal a, a corresponding b can then be computed as g (\y \rightarrow p (a,y)).

In this case, the pair function can be conceptualized as a function that constructs all possible combinations of the elements within the provided selection function and subsequently identifies the overall optimal one.

It might feel intuitive to consider the following modified pair function that seems to be more symmetric.

```
[] pair' :: J r a -> J r b -> J r (a,b) pair' f g p = (a,b) where a = f (\x -> p (x, g (\y -> p (x,y)))) b = g (\y -> p (f (\x -> p (x,y)), y))
```

However, applying this modified pair' to our previous example this results in a overall non optimal solution.

```
pair' p1 p2 pred
--> (Left,Left)
```

This illustrates how the original pair function keeps track of its first decision when determining its second element. It is noteworthy that, in the example example, achieving a satisfying outcome for both pedestrians is only possible when they consider the direction the other one is heading. The specific destination does not matter, as long as they are moving in different directions. Consequently, the original pair function can be conceived as a function that selects the optimal solution while retaining awareness of previous solutions, whereas our modified pair does not.

An issue with the original pair function might have been identified by the attentive reader. There is redundant computational work involved. Initially, all possible pairs are constructed to determine an optimal first element a, but the corresponding b that renders it an overall optimal solution is overlooked, resulting in only a being returned. Subsequently, the optimal b is recalculated based on the already determined optimal a when selecting the second element of the pair.

The primary contribution of this paper will be to illustrate and propose a solution to this inefficiency.

1.3 Sequence

The generalization of the pair function to accommodate a sequence of selection functions is the initial focus of exploration. In the context of selection functions, a **sequence** operation is introduced, capable of combining a list of selection functions into a singular selection function that, in turn, selects a list of objects:

```
[] sequence :: [J r a] -> J r [a] sequence [] p = [] sequence (e:es) p = a: as where a = e(x -> p(x : sequence es(p.(x:)))) as = sequence es(p.(a:))
```

Here, similar to the pair function, the sequence function extracts elements from the resulting list through the corresponding selection functions. This extraction is achieved by applying each function to a newly constructed property function that possesses the capability to foresee the future, thereby constructing an optimal future based on the currently examined element.

However, a notable inefficiency persists, exacerbating the issue observed in the pair function. During the determination of the first element, the sequence function calculates an optimal remainder of the list, only to overlook it and redundantly perform the same calculation for subsequent elements. This inefficiency in sequence warrants further investigation for potential optimization in subsequent sections of this research paper.

1.4 Selection monad

```
The formation of a monad within the selection functions unfolds as follows: [] (>>=) :: J r a -> (a -> J r b) -> J r b (>>=) f g p = g (f (p . flip g p)) p
```

```
[] return :: a -> J r a return x p = x
```

These definitions illustrate the monadic structure inherent in selection functions. The Haskell standard library already incorporates a built-in function for monads, referred to as sequence', defined as:

```
[] sequence' :: [J r a] -> J r [a] sequence' (ma:mas) = ma >>= \xspace \xspa
```

Notably, in the case of the selection monad, this built-in sequence' function aligns with the earlier provided sequence implementation. This inherent consistency further solidifies the monadic nature of selection functions, underscoring their alignment with established Haskell conventions.

1.5 Illustration of Sequence in the Context of Selection Functions

To ilustrate the application of the sequence function within the domain of selection functions, consider a practical scenario: the task of cracking a secret password. In this hypothetical situation, a black box predicate **p** is provided that returns **True** if the correct password is entered and **False** otherwise. Additionally, knowledge is assumed that the password is six characters long:

```
[] p :: String -> Bool p "secret" = True p _ = False Suppose access is available to a maxWith function, defined as: [] maxWith :: Ord r => [a] -> (a -> r) -> a maxWith [x] f = x maxWith (x:y:xs) f — (f x) > (f y) = maxWith (x:xs) f — otherwise = maxWith (y:xs) f
```

With these resources, a selection function denoted as **selectChar** can be constructed, which, given a predicate that evaluates each character, selects a single character satisfying the specified predicate:

```
selectChar :: J Bool Char selectChar = maxWith ['a'..'z']
```

It's worth noting that the use of maxWith is facilitated by the ordered nature of booleans in Haskell, where True is considered greater than False. Leveraging this selection function, the sequence function can be employed on a list comprising six identical copies of selectChar to successfully crack the secret password. Each instance of the selection function focuses on a specific character of the secret password:

```
sequence (replicate 6 selectChar) p
-> "secret"
```

This illustrative example not only showcases the practical application of the sequence function within the domain of selection functions but also emphasizes its utility in addressing real-world problems, such as scenarios involving password cracking. Notably, there is no need to explicitly specify a predicate for judging individual character; rather, this predicate is constructed within the monads bind definition, and its utilization is facilitated through the application of the sequence function.

Additionally, attention should be drawn to the fact that this example involves redundant calculations. Upon determining the first character of the secret password, the system neglects the prior calculation of the entire password

and recommences the calculation for subsequent characters. This inefficiency, observed in the current implementation, will be addressed through the proposal of a new type for selection functions in the subsequent section.

1.6 More efficient special K

In order to address this secific inefficiency of the selection monad with the pair and sequence function we will introduce two new variations of the selection monad. First, we will have a look at a new type K that will turn out to be isomorphic to the selection monad J. Then we will further generalise this K type to be more intuitive to work whith. It turns out that the J monad can be embedded into this generalised K type.

2 Special K

Lets consider the following type K:

```
[] type K r a = forall b. (a \rightarrow (r,b)) \rightarrow b
```

While selection functions of type J are still waiting for a predicate that is able to judge its underlaying elements, the new K type works similar. The predicate of the K type also judges its elements by turning them into r values, but further also converts the x into any y, and returns that y along with its judgement r.

```
[] pair
K :: K r a -> K r b -> K r (a,b) pair
K f g p = f (\x -> g (\y -> let (r, z) = p (x,y) in (r, (r,z))))
```

- ilustrate on an example how that is more efficient Basically because once it found a solution, the whole solution will be returned, and can be reused
- This is sequence for the new K type.

```
[] sequenceK :: [K r a] -> K r [a] sequenceK [e] p = e (x -> p [x]) sequenceK (e:es) p = e (x -> e
```

- state that it has the same efficiency advantages

2.1 Special K isomorphic to J

- Give k2j and j2k

```
[] k2j :: K r a -> J r a k2j f p = f (\x -> (p x, x))
[] j2k :: J r a -> K r a j2k f p = snd (p (f (fst . p)))
```

- intoduce free theorem for special K
- proof that they are isomorphic
- End with a final point that this is complicated to deal with! Lots of unpacking

2.2 Generalised K

- what we really want is the generalised K

[] type GK r x = forall y.
$$(x -> (r,y)) -> (r,y)$$

- give the intuitive monad definition for new K

```
[] bindGK :: GK r a -> (a -> GK r b) -> GK r b bindGK e f p = e (\x -> f x p) 
 [] returnGK :: a -> GK r a returnGK x p = p x
```

- give pair and sequence

```
[] pairGK :: GK r a -> GK r b -> GK r (a,b) pairGK f g p = f (\x -> g (\y -> p (x,y))) 
 [] sequenceGK :: [GK r a] -> GK r [a] sequenceGK [e] p = e (\x -> p [x])
```

[] sequenceGK :: [GK r a] -> GK r [a] sequenceGK [e] $p = e (\x -> p [x])$ sequenceGK (e:es) $p = e (\x -> sequenceGK es (\x -> p (x:xs)))$

- ilustrate how nice it is to deal with

2.3 Relationship to J and Special K

- Show that generalised K is an embedding

```
[] k2gk :: K r a -> GK r a k2gk f = snd . f [] gk2k :: GK r a -> K r a gk2k f p = f (\x -> let (r,y) = p x in (r, (r,y)))
```

- intoduce free theorem and precondition
- counterexamples to ilustrate what precondition means and why we want it
- introduce new theorem baced on free theorem and precondition
- calculate monad definition from k2j and j2k

3 Performance analysis

- give some perforance analysis examples that illustrate improvement

4 Related work

J was researched in the context of Sequential games, but slowly found its way to other applications

5 Outlook and future work

- Need to investigate further whats possible with the more general type
- Alpha beta pruning as next step of my work

6 Conclusion

- $-\,$ We should use generalised K is tead of J because more useful and more intuitive once understood
- performance improvements are useful
- $-\,$ monad pair and sequence implementation much more intuitive and useful

7 Appendix

Proofs!

References

Appendix