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# Converting morfisms

Given a category k we can represent its morfisms using the intent to left-compose with another morfism:

 $f:: a'k'b \text{ becomes } (\circ f):: (b'k'r) \to (a'k'r) \text{ where } r \text{ is any object of } k.$  If h is the morfism we'll compose with f then h is the continuation of f.

Creating a category based around this functor will give us our generalized RAD algorithm.

We'll begin by creating a new data type:

**newtype** Cont 
$$(k, r)$$
 a  $b = Cont ((b'k'r) \rightarrow (a'k'r))$ 

And then defining a functor from it:

cont :: Category 
$$k \Rightarrow (a' k' b) \rightarrow Cont(k,r) a b$$
  
cont  $f = Cont(\circ f)$ 

With this we can derive new categorical instances:



```
instance Category k \Rightarrow Category Cont(k, r) where
  id = Cont id
  Cont g \circ Cont f = Cont (f \circ g)
instance Monoidal k \Rightarrow Monoidal Cont (k, r) where
  Conf f x Cont g = Cont (\land \circ (f \times g) \circ unjoin)
instance Cartesian k \Rightarrow Cartesian Cont (k, r) where
  exl = Cont (\land \circ inl); exr = Cont (\land \circ inr)
  dup = Cont (jam \circ unjoin)
instance Cocartesian k \Rightarrow Cocartesian Cont (k, r) where
  inl = Cont (exl \circ unjoin); inr = Cont (exr \circ unjoin)
  jam = Cont (\land \circ dup)
instance Scalable k a \Rightarrow Scalable Cont (k, r) a where
  scale s = Cont (scale s)
```

Due to it's widespread use in ML we'll talk about a specific case of RAD: computing gradients(derivatives of functions with scalar codomains)

A vector space A over a scalar field has  $A \multimap s$  as it's dual(i.e., the linear maps of the udnerlaying field of A are it's dual). Each linear map in  $A \multimap s$  can be represented in the form of dot

Each linear map in A  $\rightarrow$  s can be represented in the form of doing u for some u :: A where

```
class HasDot(S) u where dot :: u \rightarrow (u - o s) instance HasDot(IR) IR where dot = scale instance (HasDot(S) \ a, HasDot(S) \ b) \Rightarrow HasDot(S) \ (a x b) where dot(u, v) = dot \ u \ \nabla \ dot \ v
```

Since  $Cont_k^r$  works for any type/object r we can use it with the scalar field s.

The internal representation of  $Cont_{-\infty}^s$  a b is (b - s) - (a - s) which is isomorfic to (a - s). With this in mind we can call this representation as the dual/opposite of k:

**newtype** Dual 
$$(K)$$
 a  $b = Dual(b'k'a)$ 

With this construction all we need to do to create dual representations of linear maps is to convert from  $Cont_k^S$  to  $Dual_k$  using a functor that we'll now derive:

$$asDual :: (HasDot (S) a, HasDot (S) b) \Rightarrow ((b - o s) \rightarrow (a - o s)) \rightarrow (b - o a)$$
  
  $asDual (Cont f) = Dual (onDot f)$ 

where

onDot :: 
$$(HasDot (S) \ a, HasDot (S) \ b) \Rightarrow$$
  
 $((b - o \ s) \rightarrow (a - o \ s)) \rightarrow (b - o \ a)$   
onDot  $f = dot \uparrow \{ "-1 " \} \circ f \circ dot$ 

Given this we can now derive our new categorical instances



```
instance Category k \Rightarrow Category Dual (k) where
  id = Dual id
  Dual g \circ Dual f = Dual (f \circ g)
instance Monoidal k \Rightarrow Monoidal Dual (k) where
  Dual f x Dual g = Dual(f \times g)
instance Cartesian k \Rightarrow Cartesian Dual (k) where
  exl = Dual inl; exr = Dual inr; dup = Dual iam
instance Cocartesian k \Rightarrow CocartesianDual(k) where
  inl = Dual \ exl; inr = Dual \ exr; jam = Dual \ dup
instance Scalable k \Rightarrow Scalable Dual (k) where
  scale s = Dual (scale s)
```

#### Final notes

- $\triangle$  and  $\nabla$  mutually dualize (Dual  $f \triangle$  Dual g = Dual ( $f \nabla g$ ) and Dual  $f \nabla$  Dual g = Dual ( $f \triangle g$ ))
- Using the definition from chapter 8 we can determine that the duality of a matrix corresponds to it's transposition

## Fowards-mode Automatic Differentiation(FAD)

We can use the same deductions we've done in Cont and Dual to derive a category with full right-side association, thus creating a generized FAD algorithm. This algorithm is far more apropriated for low dimention domains.

**newtype** 
$$Begin(k,r)$$
  $ab = Begin((r'k'a) \rightarrow (r'k'b))$   $begin :: Category k \Rightarrow (a'k'b) \rightarrow Begin(k,r) a b$   $begin f = Begin(f \circ)$ 

We can derive categorical instances from the functor above and we can choose r to be the scalar field s, noting that  $s \multimap a$  is isomorfic to a.