

This ain't your Daddy's Probability Monad

Virtual MuniHac

Lars Brünjes



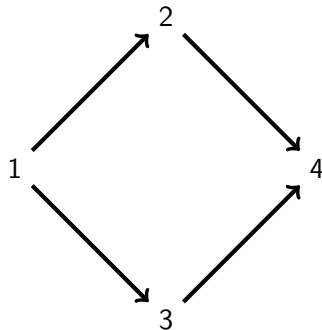
September 11, 2020

Motivation

- Traditional **probability monads** are great at modelling uncertainty.
- Things become more complicated when **time** enters the picture, especially in the presence of **concurrency**.
- Uncertainty, time and concurrency all need to be considered when trying to model the behavior of **distributed systems**.

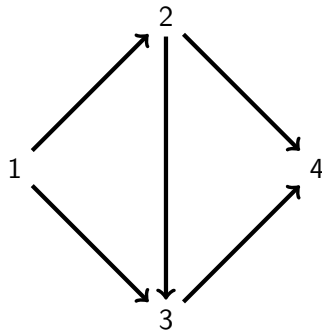
Example: The “Diamond”-Network

- We consider a network of four nodes, connected as indicated in the diagram.
- Each individual connection takes a time uniformly distributed between one second and two seconds and fails with a probability of 10%.
- What is the probability for a signal originating in node 1 to reach node 4? How long will it take?



Example: The “Diamond”-Network

- We consider a network of four nodes, connected as indicated in the diagram.
- Each individual connection takes a time uniformly distributed between one second and two seconds and fails with a probability of 10%.
- What is the probability for a signal originating in node 1 to reach node 4? How long will it take?
- How do time and probability change when an extra edge is added?



Modelling Uncertainty — MonadProb

```
type Prob p = (Ord p, Fractional p, Real p)
```

```
fromDouble :: (Fractional a, Real a) => Double -> a  
fromDouble = fromRational . toRational
```

Modelling Uncertainty — MonadProb

```
type Prob p = (Ord p, Fractional p, Real p)
```

```
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fromDouble = fromRational . toRational
```

```
class (Prob p, Monad m) => MonadProb p m | m -> p where
```

```
  coin :: p -> m Bool
```

```
  pick :: NonEmpty a -> m a
```

```
  pick (x :| []) = return x
```

```
  pick (x :| y : ys) = do
```

```
    let len = length ys
```

```
        p = recip $ fromIntegral $ len + 2
```

```
    b <- coin p
```

```
    if b then return x
```

```
        else pick $ y :| ys
```

```
unsafePick :: MonadProb p m => [a] -> m a
```

```
unsafePick = pick . fromList
```

Example: Throwing Dice

```
die :: MonadProb p m => m Int
die = pick $ 1 :| [2 .. 6]
```

```
dice :: MonadProb p m => Int -> m Int
dice n
  | n <= 0    = return 0
  | otherwise = sum <$> replicateM n die
```



Example: Monty Hall

```
data Prize = Goat | Car deriving (Show, Read, Eq, Ord)
```

```
data Strategy = Stay | Change deriving (Show, Read, Eq, Ord)
```

```
monty :: MonadProb p m => Strategy -> m Prize
```

```
monty s = do
```

```
  let doors = 0 :| [1, 2]
```

```
  carDoor  <- pick doors
```

```
  let prizes = (\i -> if i == carDoor then Car else Goat) <$> doors
```

```
  guess    <- pick doors
```

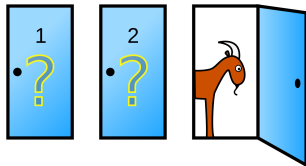
```
  openedDoor <- unsafePick $ filter (\i -> i /= guess && prizes !! i == Goat) doors
```

```
  let guess' = case s of
```

```
    Stay    -> guess
```

```
    Change -> head $ filter (\i -> i /= guess && i /= openedDoor) doors
```

```
  return $ prizes !! guess'
```



First Implementation: Sampling

```
newtype ProbS p m a = PS {runProbS :: m a }  
  deriving (Functor, Applicative, Monad, MonadRandom)  
  
instance (Prob p, MonadRandom m) => MonadProb p (ProbS p m) where  
  coin p = do  
    x <- fromDouble <$> getRandomR (0, 1)  
    return $ x <= p
```

Trying Sampling

```
diceS :: Int -> IO [Int]
```

```
diceS c = runProbS $ replicateM c (dice 2 :: ProbS Double IO Int)
```

```
->>> diceS 40
```

```
[6,6,8,9,6,7,8,5,7,5,9,5,5,6,11,8,3,7,9,8,10,9,6,9,10,8,9,4,3,10,11,2,7,11,6,6,4,6,7,7]
```

Trying Sampling

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diceS :: Int -> IO [Int]
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```
[6,6,8,9,6,7,8,5,7,5,9,5,5,6,11,8,3,7,9,8,10,9,6,9,10,8,9,4,3,10,11,2,7,11,6,6,4,6,7,7]
```

```
montyS :: Int -> Strategy -> IO [Prize]
```

```
montyS c s = runProbS $ replicateM c (monty s :: ProbS Double IO Prize)
```

```
->>> montyS 15 Stay
```

```
[Goat,Goat,Goat,Car,Goat,Car,Goat,Goat,Goat,Car,Goat,Goat,Goat,Goat,Car]
```

```
>>> montyS 15 Change
```

```
[Goat,Car,Car,Car,Goat,Goat,Car,Car,Goat,Car,Goat,Goat,Car,Car,Car]
```

Second Implementation: Exact Distribution

```
newtype ProbL p a = PL {runProbL :: [(a, p)]}  
  deriving (Show, Read, Eq, Ord, Functor)
```

- A pair (a, p) means that the computation will have result a with probability p .
- The sum over all p is 1 (not expressed by the type).
- “Morally” we would like `Map a p`, but we do not have an `Ord`-instance for all a .

Second Implementation: Exact Distribution — Instances

```
instance Prob p => Applicative (ProbL p) where
  pure = return
  (<*>) = ap
```

```
instance Prob p => Monad (ProbL p) where

  return a = PL [(a, 1)]

  PL aps >>= cont = PL $ do
    (a, p) <- aps
    (b, q) <- runProbL $ cont a
    return (b, p * q)
```

```
instance Prob p => MonadProb p (ProbL p) where
  coin p
    | p <= 0    = return False
    | p >= 1    = return True
    | otherwise = PL [(True, p), (False, 1 - p)]
```

```
probLOrd :: (Prob p, Ord a) => ProbL p a -> Map a p
probLOrd = foldl' f Map.empty . runProbL
  where
    f m (a, p) = Map.insertWith (+) a p m
```

Trying Exact Distribution

```
diceL :: [(Int, Rational)]  
diceL = Map.toList $ probLOrd $ dice 2
```

```
-->>> diceL  
[(2,1 % 36),(3,1 % 18),(4,1 % 12),(5,1 % 9),(6,5 % 36),  
 (7,1 % 6),(8,5 % 36),(9,1 % 9),(10,1 % 12),(11,1 % 18),(12,1 % 36)]
```

Trying Exact Distribution

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diceL :: [(Int, Rational)]  
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-->>> diceL  
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 (7,1 % 6),(8,5 % 36),(9,1 % 9),(10,1 % 12),(11,1 % 18),(12,1 % 36)]
```

```
montyL :: Strategy -> [(Prize, Rational)]  
montyL = Map.toList . probLOrd . monty
```

```
-->>> montyL Stay  
[(Goat,2 % 3),(Car,1 % 3)]  
>>> montyL Change  
[(Goat,1 % 3),(Car,2 % 3)]
```

Adding Delays

```
type Time t = (Fractional t, Ord t)
```

```
class ( MonadError () m  
    , MonadProb p m  
    , Time t  
    ) => MonadDelay p t m | m -> p t where  
    delay :: t -> m ()
```

```
absurd :: MonadDelay p t m => m a  
absurd = throwError ()
```

```
catch :: MonadDelay p t m => m a -> m a -> m a  
catch m h = catchError m $ const h
```


Approximating Uniform Delays

```
uniform :: Ord t => MonadDelay p t m => t -> t -> Int -> m ()
uniform a b n
  | b < a      = absurd
  | b == a     = delay a
  | otherwise  = do
    let d = (b - a) / fromIntegral n
    t <- pick $ a :| [a + d * fromIntegral i | i <- [1 .. n]]
    delay t
```

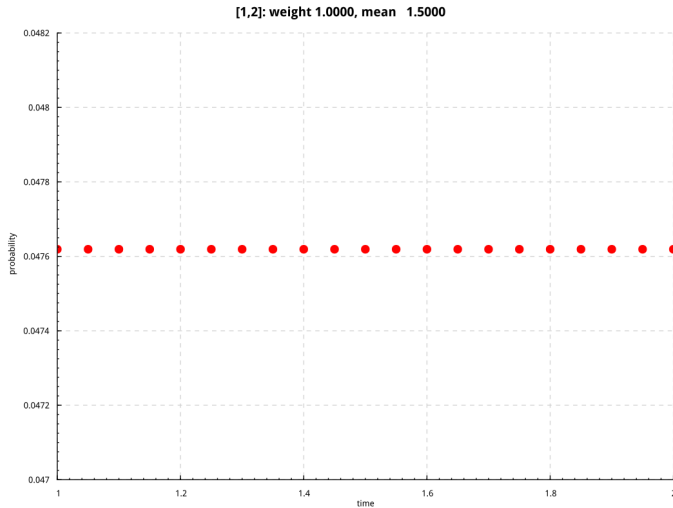
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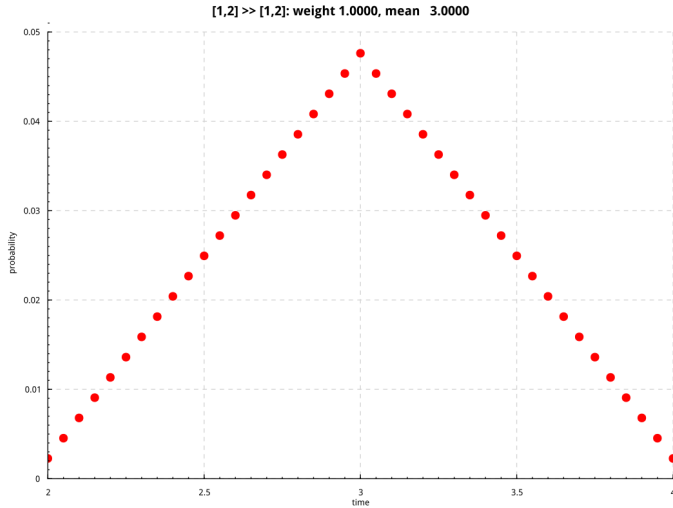
Remark

We could consider generalized delays instead to include “proper” uniform distributions, but they make implementing more difficult.

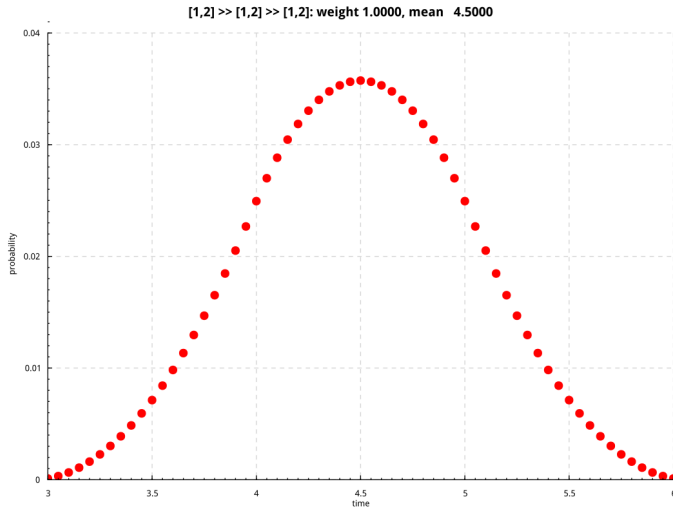
Example: $d = \text{uniform } 1 \ 2 \ 20$



Example: $d \gg d$



Example: $d \gg d \gg d$



Just a Standard Transformer Stack

```
newtype DelayT p t m a = DT (WriterT (Sum t) (ExceptT () m) a)
deriving (Functor, Applicative, Monad, MonadError (), MonadWriter (Sum t))
```

```
instance Time t => MonadTrans (DelayT p t) where
    lift = DT . lift . lift
```

```
instance (Time t, MonadProb p m) => MonadProb p (DelayT p t m) where
    coin = lift . coin
```

```
instance (Time t, MonadProb p m) => MonadDelay p t (DelayT p t m) where
    delay = tell . Sum
```

Adding Concurrency

```
class MonadDelay p t m => MonadRace p t m | m -> p t where  
  race :: m a -> m b -> m (Either (a, m b) (m a, b))
```

Racing

We can “race” two computations and — after some time — get back the result of one and what remains of the other.

First-to-Finish Synchronization

```
ftf :: MonadRace p t m => [m a] -> m a
ftf [] = absurd
ftf (ma : mas) = either fst snd <$> race ma (ftf mas)
```


Last-to-Finish Synchronization

```
ltf :: MonadRace p t m => [m a] -> m [a]
ltf [] = return []
ltf (ma : mas) = do
  e <- race ma $ ltf mas
  case e of
    Left (a, m)  -> (a :) <$> m
    Right (m, xs) -> m >>= \a -> return (a : xs)
```

Last-to-Finish Synchronization

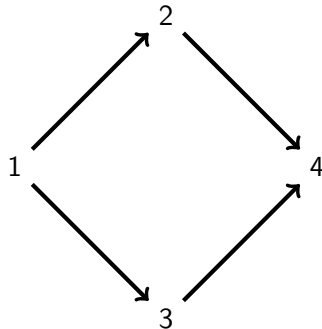
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ltf (ma : mas) = do
  e <- race ma $ ltf mas
  case e of
    Left (a, m)  -> (a :) <$> m
    Right (m, xs) -> m >>= \a -> return (a : xs)
```

Remark

We could use `ftf` and `ltf` as primitives instead of `race`. This would be strictly less powerfull, but on the other hand would allow for more complicated delays (not just discrete ones).

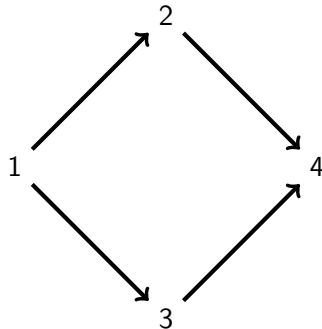
Implementing the “Diamond”-Network

```
-- one edge  
d :: (Ord t, Num t, MonadRace p t m) => m ()  
d = do  
    b <- coin 0.9  
    if b then uniform 1 2 20 else absurd
```



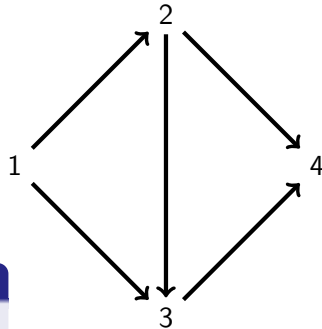
Implementing the “Diamond”-Network

```
diamond1 :: MonadRace p t m => m ()  
diamond1 = do  
  let d12 = d; d13 = d; d24 = d; d34 = d  
  ftf [d12 >> d24, d13 >> d34]
```



Implementing the “Diamond”-Network

```
diamond2 :: MonadRace p t m => m ()  
diamond2 = do  
  let d12 = d; d13 = d; d23 = d; d24 = d; d34 = d  
  e <- race d12 d13  
  case e of  
    Left  ((()), r13)  -> ftf [d24, ftf [d23, r13] >> d34]  
    Right ((r12, ())) -> ftf [d34, r12 >> d24]
```



Remark

The full power of `race` is needed — `ftf` alone won't do!

First Implementation of MonadRace: Sampling

```
newtype RaceS p t m a = RS {runRS :: m (Maybe (a, t))}  
    deriving Functor
```

- A Nothing return-value corresponds to failure.
- A Just (a, t) return-value indicates result a after time t.

First Implementation of MonadRace: Sampling — Monad

```
instance (Time t, MonadRandom m) => Applicative (RaceS p t m) where
    pure = return
    (<*>) = ap
```

```
instance (Time t, MonadRandom m) => Monad (RaceS p t m) where
    RS x >>= cont = RS $ do
        mat <- x
        case mat of
            Nothing    -> return Nothing
            Just (a, t) -> do
                mbs <- runRS $ cont a
                case mbs of
                    Nothing    -> return Nothing
                    Just (b, s) -> return $ Just (b, t + s)
```

First Implementation of MonadRace: Sampling — MonadError

```
instance (Time t, MonadRandom m) => MonadError () (RaceS p t m) where
```

```
  throwError () = RS $ return Nothing
```

```
  catchError (RS x) h = RS $ do
```

```
    mat <- x
```

```
    case mat of
```

```
      Nothing -> runRS $ h ()
```

```
      Just _   -> return mat
```


First Implementation of MonadRace: Sampling — MonadProb

```
instance (Prob p, Time t, MonadRandom m) => MonadProb p (RaceS p t m) where
  coin p = RS $ do
    x <- fromDouble <$> getRandomR (0, 1)
    return $ Just (x <= p, 0)
```

First Implementation of MonadRace: Sampling — MonadDelay

```
instance (Prob p, Time t, MonadRandom m) => MonadDelay p t (RaceS p t m) where
  delay t = RS $ return $ Just ((), t)
```

First Implementation of MonadRace: Sampling — MonadRace

```
instance (Prob p, Time t, MonadRandom m) => MonadRace p t (RaceS p t m) where
  race (RS x) (RS y) = RS $ do
    mat <- x
    mbs <- y
    case (mat, mbs) of
      (Nothing,    Nothing)    -> return Nothing
      (Just (a, t), Nothing)    -> return $ Just (Left (a, absurd), t)
      (Nothing,    Just (b, s)) -> return $ Just (Right (absurd, b), s)
      (Just (a, t), Just (b, s))
        | t <= s                -> return $ Just (Left (a, delay (s - t) >> return b), t)
        | otherwise             -> return $ Just (Right (delay (t - s) >> return a, b), s)
```

Pros and Cons of Sampling

- Sampling is straight forward and efficient.
- We could even support more sophisticated delays, like “proper” uniform delays easily.
- On the other hand, it would be nice to get **exact** results (at least for small examples).

Second Implementation of MonadRace: Exact Distribution

```
newtype RaceL p t a = RL {runRL :: [(t, p, a)]}  
  deriving (Show, Read, Eq, Ord, Functor)
```

- A triple (t, p, a) corresponds to result a being obtained after time t with probability p .
- The sum of all p in the list, the **weight**, can be strictly smaller than 1, in which case there is a positive probability that the computation **fails**.
- “Morally” we would like $\text{Map } a \ (\text{Map } t \ p)$, but again we do not have an `Ord`-instance for all a .

Second Implementation of MonadRace: Exact Distribution — MonadError

```
instance (Prob p, Time t) => Monad (RaceL p t) where
  return a = RL [(0, 1, a)]
  RL xs >>= cont = RL $ do
    (ta, pa, a) <- xs
    (tb, pb, b) <- runRL $ cont a
    return (ta + tb, pa * pb, b)
```

```
instance (Prob p, Time t) => MonadError () (RaceL p t) where
  throwError () = RL []
  catchError (RL []) h = h ()
  catchError m _ = m
```

Second Implementation of MonadRace: Exact Distribution — MonadProb

```
instance (Prob p, Time t) => MonadProb p (RaceL p t) where
  coin p
    | p <= 0    = return False
    | p >= 1    = return True
    | otherwise = RL [(0, p, True), (0, 1 - p, False)]
```

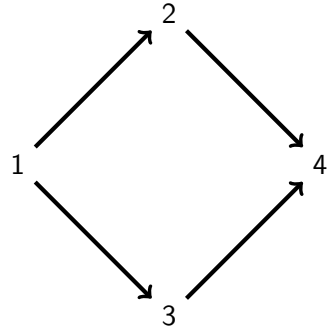
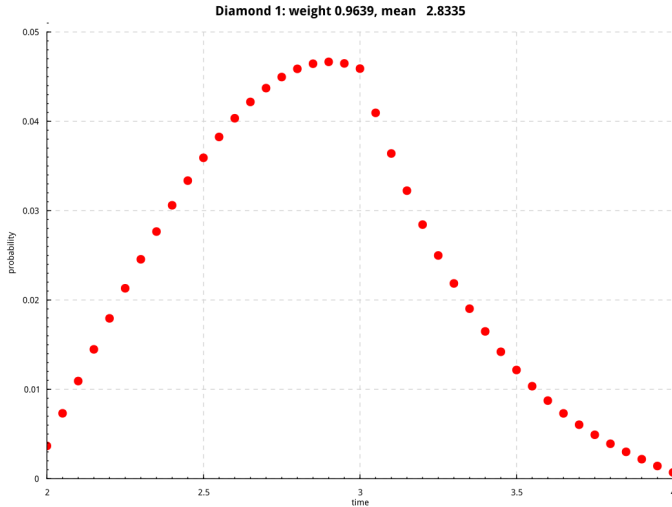
Second Implementation of MonadRace: Exact Distribution — MonadDelay

```
instance (Prob p, Time t) => MonadDelay p t (RaceL p t) where
  delay t = RL [(t, 1, ())]
```

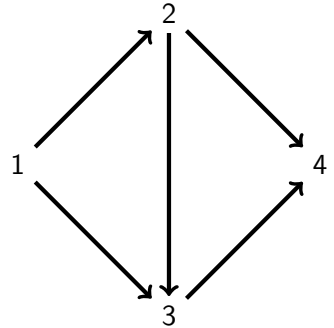
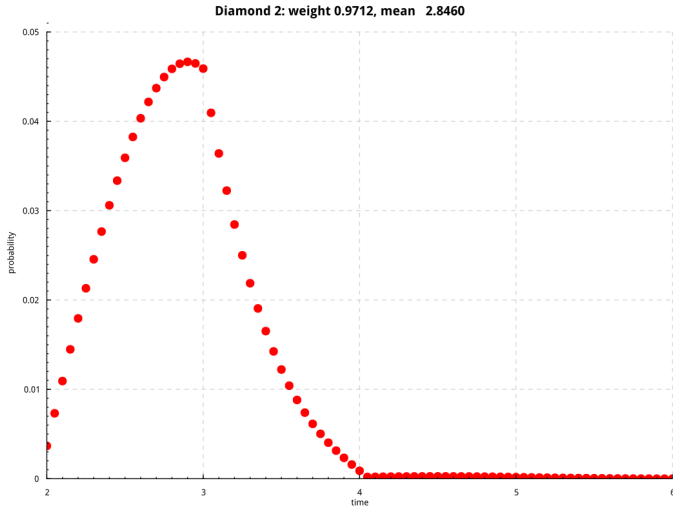

Second Implementation of MonadRace: Exact Distribution — MonadRace

```
instance (Prob p, Time t) => MonadRace p t (RaceL p t) where
  race (RL xs) (RL ys) = RL $ do
    let fx = max 0 $ min 1 $ 1 - weight xs
        fy = max 0 $ min 1 $ 1 - weight ys
        za = if fy > 0 then [(ta, pa * fy, Left (a, absurd)) | (ta, pa, a) <- xs] else []
        zb = if fx > 0 then [(tb, pb * fx, Right (absurd, b)) | (tb, pb, b) <- ys] else []
        zab = do
          (ta, pa, a) <- xs
          (tb, pb, b) <- ys
          return $ if ta <= tb
            then (ta, pa * pb, Left (a, RL [(tb - ta, 1, b)]))
            else (tb, pa * pb, Right (RL [(ta - tb, 1, a)], b))
    za ++ zb ++ zab
  where
    weight :: [(t, p, c)] -> p
    weight ws = sum [p | (_, p, _) <- ws]
```

Example: The “Diamond”-Network



Example: The “Diamond”-Network



Thank you for your attention!



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