



Faculty of Science



Monte-Carlo based Pricing in Haskell

(sliced out of other presentations)

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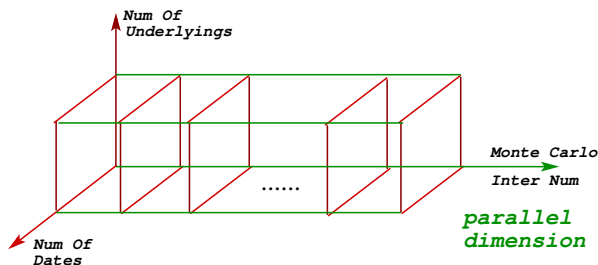
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Bird's Eye View



- Generate independent Sobol sequences (uniform in $[0, 1)$)
- Transform into Gaussian distribution $(-\infty, \infty)$ (numerical \int)
- Brownian bridge for all dates, for each underlying
- Compute trajectory using Black-Scholes
- Monte-Carlo aggregation with product's payoff and discount



Bird's Eye View

Haskell Code

```
mc_pricing n = let conf = init_sobol_orig n
                gains  = map ( payoff      conf
                             . black_scholes conf
                             . brownian_bridge conf
                             . gaussian
                             . sobolInd      conf ) [1..n]
                price  = sum gains
                factor = 1 / (fromIntegral N)
            in  factor * price
```

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Functional Basic Blocks

Haskell Code Inlined

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```

Functional Basic Blocks Types

```
sobolInd      :: Pricing_Data -> Index -> [Real]
-- sobolRec   :: Pricing_Data -> [Elem] -> Index -> [Real]
gaussian      :: [Real] -> [Real]
-----
brownian_bridge :: Pricing_Data -> [Real] -> [[Real]]
black_scholes   :: Pricing_Data -> [[Real]] -> [[Real]]
payoff          :: Pricing_Data -> [[Real]] -> Real
```

$$[0, 1)^{u \cdot d}$$

$$[0, 1)^{u \cdot d}$$

$$\mathbb{R}^{u \cdot d}$$

$$\mathbb{R}^{u \times d}$$

$$\mathbb{R}^{u \times d}$$

$$\mathbb{R}$$


Computing a New Trajectory with Black-Scholes

Black-Scholes

```
black_scholes :: Pricing_Data -> [[SpecReal]] -> [[SpecReal]]
black_scholes c = mkPrices c . correlate_deltas c . mkDeltas
```

Compute deltas:

$$\delta_{i,j} = r_{i,j} - r_{i-1,j} \quad \forall_{i>0}, \quad \delta_{0,j} = r_{0,j}$$

```
mkDeltas :: Num r => [[r]] -> [[r]]
mkDeltas rows@(row1:restrows) = row1 : zipWith (zipWith (-)) restrows rows
```

Correlate delta matrix using C matrix:

$$\delta_{i,j}^c = \sum_{k=0}^j \delta_{i,k} \cdot c_{j,k}$$

```
correlate_deltas :: Pricing_Data -> [[SpecReal]] -> [[SpecReal]]
correlate_deltas Pricing_Data{..} zds
  = map (\zr -> map (sum . zipWith (*) zr) md_c) zds
```

Compute Prices with Black-Scholes:

$$s_{i,j} = s_{0,j} \cdot e^{\sum \delta_{k,j}^c \cdot \sigma_{k,j} + dr_{k,j}}$$

```
mkPrices :: Pricing_Data -> [[SpecReal]] -> [[SpecReal]]
mkPrices Pricing_Data{..} noises
  = let e_rows = map (map exp) -- dc[l][j]*vol[l][j] + dr[l][j]
    [ zipWith (+) (zipWith (*) n_row vol_row) dr_row
      | (n_row, vol_row, dr_row) <- zip3 noises md_vols md_drifts ]
  in tail (scanl (zipWith (*)) md_starts e_rows)
```



Black-Scholes: C++

C++ Code

```
for (i = n_und * n_dates - 1; i >= n_und; i--) z[i] -= z[i-n_und];

for (int i = 0; i < n_dates; i++) {
  for (int j = 0; j < n_und; j++) {
    double temp = 0.0;
    k = n_und * i + j;
    for (l = 0; l <= j; l++)
      temp += md_c[n_und * j + l] * md_z[n_und * i + l];
    temp = exp(temp * md_vols[k] + md_drift[k]);
    trajectory[k] = trajectory[k - n_und] * temp;
  }
}
```

- Largely uses **destructive updates**, reusing same array.
- **Deltas**: traversed from the back ("antidependence").
- **Prefix Sum**: appears as flow-dependency (constant distance)
- Difficult (for compiler) to recognize parallelism opportunities.



Performance Comparison & Discussion

	Europ.Call 10^7 samples
Haskell, first version	21.20 sec
Using lists everywhere	16.02 sec
Using arrays for Sobol direction vectors	11.88 sec
Using arrays and recurrence	5.98 sec
C++ version	1.09 sec

(Intel Core i7 2.30GHz)

The jump between 21.2 to 16 sec was due to hoisting constant data-structures at global level.



Sobol Quasi-Random Numbers

Sobol random number generator: generate a sequence of values $\{x^1, x^2, \dots, x^N\}$ with low discrepancy $O(\log N)$ over interval $[0, 1)$.

- Chose $P \in \mathbb{Z}_2[X]$ primitive polynomial of degree d :

$$P = X^d + a_1 X^{d-1} + \dots + a_{d-1} X + 1,$$
- Compute the direction vectors:

$$m_i = 2a_1 m_{i-1} \oplus \dots \oplus 2^{d-1} a_{d-1} m_{i-d+1} \oplus 2^d m_{i-d} \oplus m_{i-d}$$
- To generate the n 'th random number, $n = \dots b_3 b_2 b_1$,

$$x^n = b_1 v_1 \oplus b_2 v_2 \oplus \dots, \text{ where } v_i = m_i / 2^i \in [0, 1)$$
- Using Gray code: $\dots g_3 g_2 g_1 = \dots b_3 b_2 b_1 \oplus \dots b_4 b_3 b_2$

$$x^{n+1} = x^n \oplus v_c, \text{ where } g_c \text{ is the rightmost 0 in } n\text{'s Gray-code rep.}$$

... recursion, can be used for optimisations
- Generalization to S dimensions under $O(\log^S N)$ discrepancy.



Recursive Sobol for Optimization

Independent Sobol Formula

vs.

Recurrent Sobol Formula

```
sobolInd :: Pricing_Data ->
          Index -> [ Elem ]
sobolInd Pricing_Data{..} i =
  let biton = testBit (grayCode i)
      inds  = filter biton
              [0..SBC-1]
      xorVs vs = foldl' xor 0
                  [vs!j | j<-inds]
  in map xorVs sobol_dirVs
map (sobolInd conf) [1..n]
```

```
sobolRec :: Pricing_Data ->
          [Elem] -> Index -> [Elem]
sobolRec Pricing_Data{..} prev i =
  let bit    = pos_least_sig_0 i
      dirVs  = [ vs!bit |
                  vs<-sobol_dirVs ]
  in zipWith xor prev dirVs
sobolRecMap conf (1,u) =
  let a = sobolInd conf 1
  in scanl (sobolRec conf) a [1..u-1]
sobolRecMap conf (1,n)
```

- Independent: *more computation*, *but* embarrassingly parallel alg.
- Recurrent: *less computation*, *but* $\log(N)$ -depth parallel alg.
- Can we have the benefits of both?



Algorithm-Level Optimizations: Sobol Generator

Optimize Parallelism Via Serial Chunking

```
-- Incorrect for arbitrary l      mapChunkList :: ([b] -> [a]) -> [b] -> Int -> [a]
sobelRecMap conf l =            mapChunkList fun l c =
  let a = sobolInd conf         let sliced :: [[b]]
                                sliced = chunkIt c l
                                out      = map fun sliced
  in scanl (sobelRec conf) a    in foldl (++) [] (map fun sliced)
    (map (+ (-1)) (tail l))
```

- Can write `mapChunkList` to chunk-map a function on an arbitrary list,
- however `sobelRecMap` is correct only when `l` contains consecutively increasing integers (hence incorrect for arbitrary `l`).



Optimization

Code Ensuring the Invariant

```
sobolRecMap conf (1,u) =
  let a = sobolInd conf 1
  in scanl
      (sobolRec conf)
      a      [1..u-1]
```

Embarrassingly parallel on N/TILE procs: the **expensive** `sobolInd` is amortized against `TILE` (fast) `sobolRec` computations.

Speedup: 1.8x to 3.75x.

User Expresses Algorithmic Invariants; Compiler Optimizes.

```
-- INVAR: sobolInd c n+1 == sobolRec c (sobolInd c n) n
sobolRec :: Pricing_Data -> [Elem] -> Index -> [Elem]
map (sobolInd conf) [1..n]
```

```
-- Compiler Explores The Rich Optimization Space
sobolGen conf n = case (cost_model conf) of
  1 -> doallChunk (sobolRecMap conf) n tile
  2 -> map (sobolInd conf) [1..n]
  3 -> sobolRecMap conf (1,n)
```

