

Master's thesis defence

Option pricing using functional data-parallel languages

Philip Carlsen Martin Dybdal

Computer Science
University of Copenhagen

22. March 2013

Overview

- Data-parallel language survey
 - ▶ Comparison through implementation of option pricing algorithms
 - ▶ Identification of shortcomings

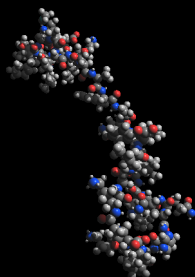



Overview

- Data-parallel language survey
 - ▶ Comparison through implementation of option pricing algorithms
 - ▶ Identification of shortcomings
- Proposed extensions
 - ▶ Not completely implemented

Data-parallel problems: Finance



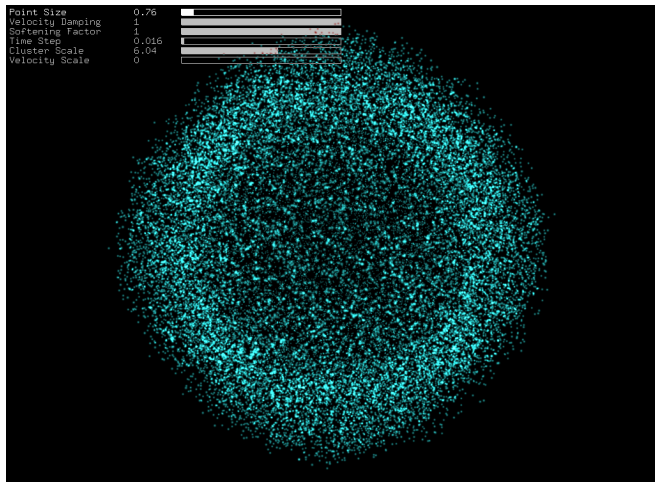
Data-parallel problems: Bioinformatics



Current Work Unit		Donor	
Name:	p5000_supervil	Name:	Jensen Huang
Progress:	2999698 / 5000000 = 60%	Team:	Team WhoopAss
Performance:	317 ns / day	Hardware:	GeForce 9600 GT
Time Left:	00d:00h:18m:10s		

Protein Folding

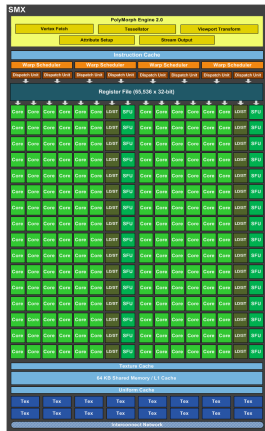
Data-parallel problems: Scientific simulations



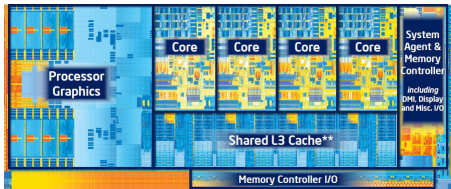
N-body simulation

GPUs vs. CPUs

NVIDIA GK104 GPU
Multiprocessor



Intel Ivy Bridge CPU



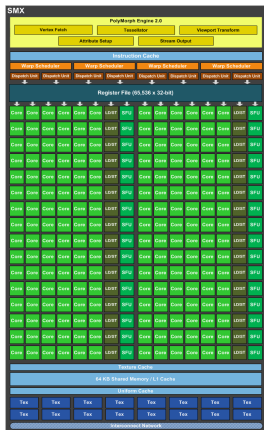
- Computation vs. caching
- High throughput vs. low latency

GPU programming

NVIDIA GK104 GPU Multiprocessor

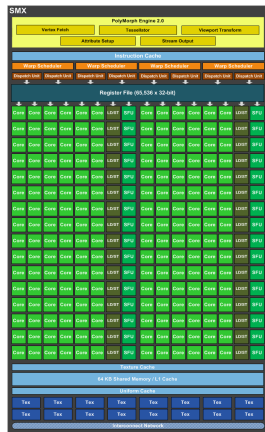
Things to be aware of:

- Memory access coordination



GPU programming

NVIDIA GK104 GPU Multiprocessor

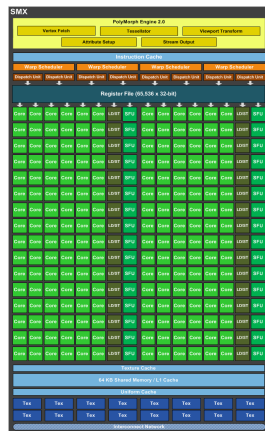


Things to be aware of:

- Memory access coordination
- Avoid memory accesses (perform program fusion)

GPU programming

NVIDIA GK104 GPU Multiprocessor



Things to be aware of:

- Memory access coordination
- Avoid memory accesses (perform program fusion)
- Schedule enough work

GPU programming languages

What we want:

- Avoid manual memory management
- Avoid manual fusion
- Fusion with library code
- Good abstractions
- Hardware independent

Data-parallel programming languages

- Feldspar
- Obsidian (GPU)
- Data-parallel Haskell
- Copperhead (GPU)
- Nikola (GPU)
- Accelerate (GPU)
- Repa
- Data.Vector

Later we discovered: Theano, Bohrium, CnC-CUDA and R+GPU

Data-parallel programming languages

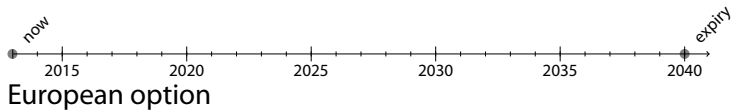
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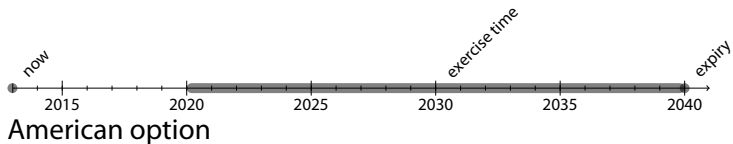
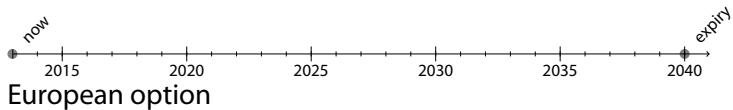
Options

See blackboard

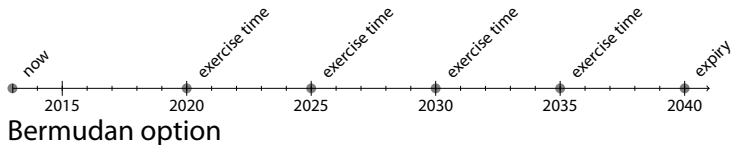
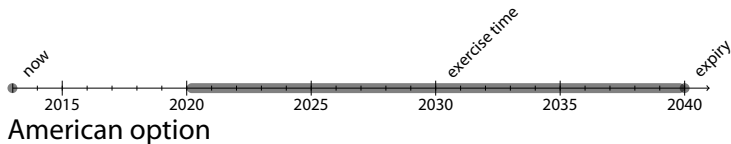
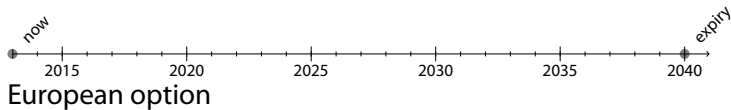
Option styles



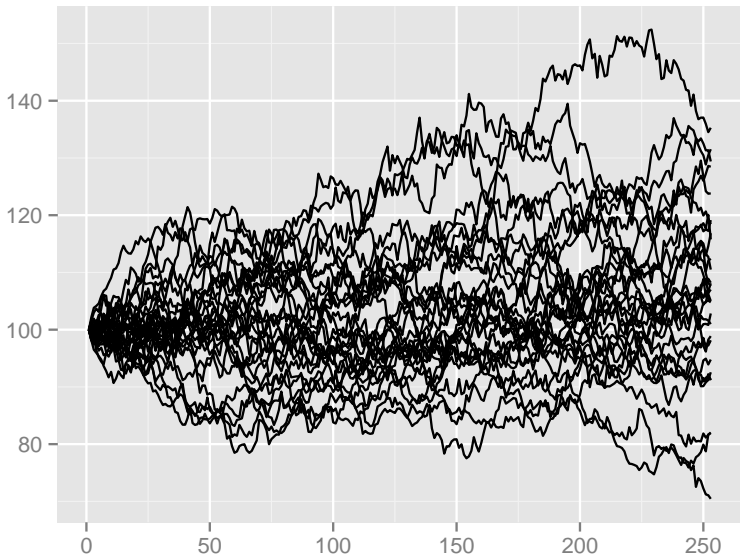
Option styles



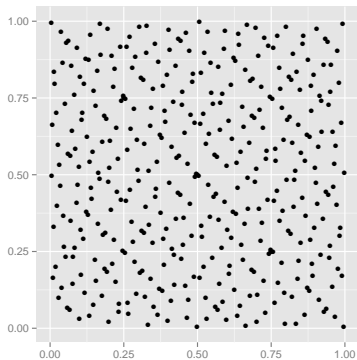
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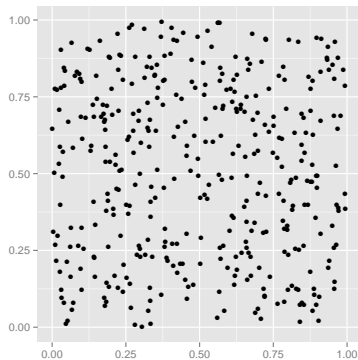
Option pricing: Least-Squares Monte Carlo



Sobol-sequence generation



2D Sobol-sequence



2D Mersenne Twister Sequence

Sobol-sequence generation

What we want:

SobolInductive $v\ i = \mathbf{reduce}\ (\oplus)\ 0\ (\mathbf{zipWith}\ (\cdot)\ v\ (\mathbf{ToBitVector}\ i))$

$$\left(\left(\begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_{32} \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 1 \\ \vdots \\ 0 \end{pmatrix} \right) \right) \overset{\oplus}{\downarrow} = x_3$$

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$$\text{SobolSequence1D } m \ v = \mathbf{parmap} \ (\text{SobolInductive } v) \ [1..m]$$

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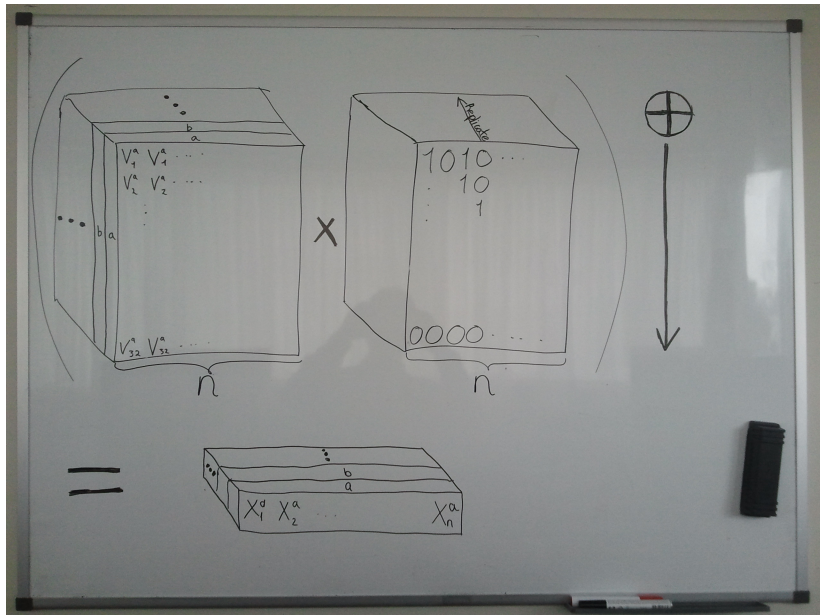
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Sobol-sequence generation



Implementation observations

- Rely on fusion to avoid memory overhead

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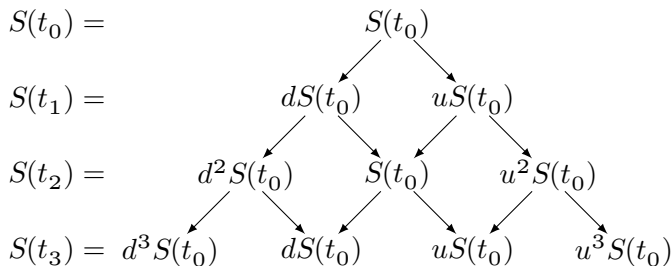
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Implementation observations

- Rely on fusion to avoid memory overhead
- Incompositional
- Thus, library development seems infeasible

Option pricing: Binomial method

$$S(t + \Delta t) = \begin{cases} S(t)u & \text{with probability } q \\ S(t)d & \text{with probability } 1 - q \end{cases}$$



Binomial lattice for three periods ($u \cdot d = 1$)

Implementation observations

Accelerate and Nikola:

- Synchronization has to happen across all blocks.
- We cannot force the use of just a single block.
- We cannot reuse already allocated memory.
- Writing portfolio pricer requires irregular arrays.

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- Writing portfolio pricer requires irregular arrays.

BinomialPortfolio = **parmap** (λ opt. **foldl** (prev opt) (final opt) [n..1])

Nested array operations

The same operations can execute in several ways:

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- **mapP** - independent parallel threads
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We can now write:

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$\text{SobolSequence1D } m\ v = \text{mapP } (\text{SobolInductive } v) \ [1..m]$

$\text{SobolSequenceND } m\ vs = \text{mapSeqPar } (\text{SobolSequence-1D } m) \ vs$

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Block level synchronization

Several additional levels, depending on the architecture:

- **mapB** - independent parallel threads using only in a single block
- **mapSeqParB** - sequential map with device synchronization between each map.

BinPortfolio = **map (foldl (map ...) (map ...) [n..1])**

BinPortfolio = **mapP (foldlSeqParB (mapB ...) (mapB ...) [n..1])**

Ordering

- **S** can occur within **P** or **B** operations
- **B** can occur within **SeqPar**, **P** or **SeqParB** operations
- **SeqParB** can occur within **SeqPar**, **P** or **SeqParB** operations
- **P** can occur within **SeqPar** operations
- **SeqPar** can occur within **SeqPar**

Sobol-sequences take #2

Recursive formulation: $x_{i+1} = x_i \oplus v_{\text{lsb}(i)}$

unfold :: (Int → a → a) → Int → a → [a]

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Future work

- Implementation and theoretical development
 - ▶ Will this work out in practice?
 - ▶ How should memory be managed?
 - ▶ How are irregular arrays handled?
 - ▶ Can we do this in the type system?
 - ▶ **mapP**'s and **foldlS**'s as a target language in other situations

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 - ▶ **mapP**'s and **foldlS**'s as a target language in other situations
- Extend the survey
 - ▶ Test additional languages
 - ▶ Implement complete LSM
 - ▶ Evaluate implementations of additional algorithms