

Monte Carlo simulation of Hull-White model and sensitivities computation

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Project Overview: The Hull-White Model

1. Estimation of $P(0, T)$ and $f(0, T)$

$$dr(t) = [\theta(t) - ar(t)]dt + \sigma dW_t$$

- ▶ **Discrete simulation:** $r(t) = m_{s,t} + \Sigma_{s,t} G$
 - ▶ $m_{s,t}$ is the deterministic drift given $r(s)$;
 - ▶ $\Sigma_{s,t}$ is the accumulated volatility of $r(t)$ and $G \sim \mathcal{N}(0, 1)$.
- ▶ **Zero coupon bond price ($P(0, T)$) and forward rate ($f(0, T)$):** estimated by Monte Carlo from interest rate r .
- ▶ **Drift ($\theta(t)$):** term calibrated to fit the initial market data.

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2. Estimation of $\theta(t)$ and $\text{ZBC}(5, 10, e^{-0.1})$

From $f(0, T)$ recover the drift $\theta(t)$ and from analytical expressions of $P(t, T)$ compute the value of the European call option $\text{ZBC}(S_1, S_2, K)$, discounted expected payoff given S_1, S_2, K .

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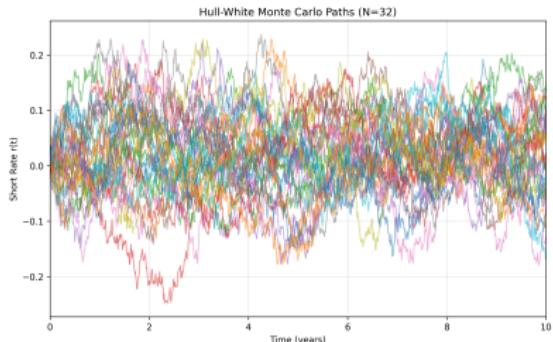
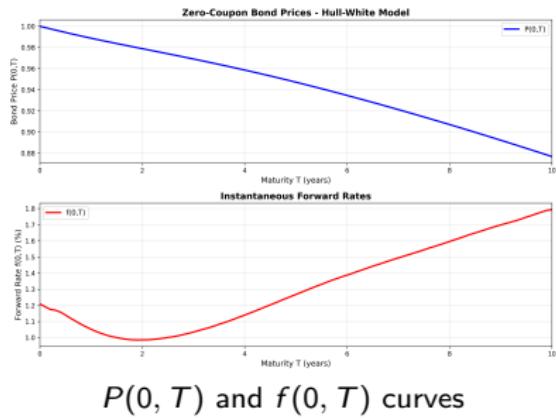
3. Estimation of $\partial_\sigma \text{ZBC}(S_1, S_2, K)$

The sensitivity of $\text{ZBC}(S_1, S_2, K)$ w.r.t. the parameter σ determining the volatility of $r(t)$, estimated by Monte Carlo.

1. Monte Carlo Estimation of $P(0, T)$ and $f(0, T)$

Computed by 2^{21} paths:

- ▶ Generated bond prices $P(0, T)$ and forward rates $f(0, T)$ using 1000 timesteps between $[0, T]$.
- ▶ Simulation time = 5.36 ms
- ▶ Throughput = 391 M paths/sec
- ▶ $f(0, 0) = 1.21\%$ (Matches initial short rate r_0)



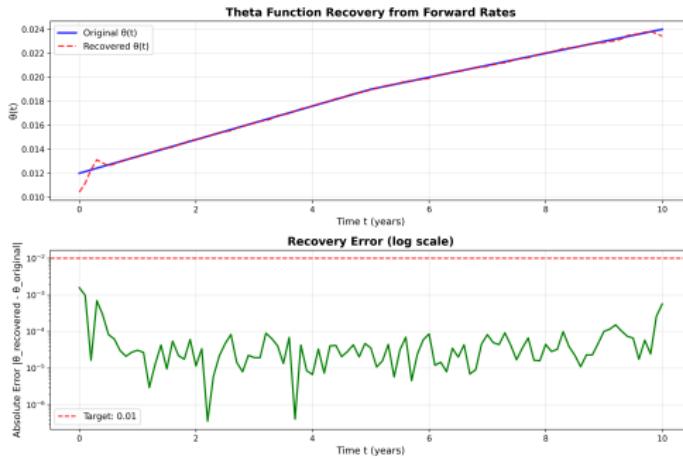
Sample Short Rate Paths $r(t)$

2a. $\theta(t)$ Calibration

- ▶ A thread for each t in the discretized interval $[0, T]$.
- ▶ Recover a given piecewise linear expression of $\theta(t)$ by

$$\theta(t) = \frac{\partial f(0, T)}{\partial T} + af(0, T) + \frac{\sigma^2(1 - e^{-2aT})}{2a}$$

- ▶ Mean error = $2.49 * 10^{-4}$



Recovered $\theta(t)$ and its error

2b. Monte Carlo Estimation of $\text{ZBC}(5, 10, e^{-0.1})$

Computed by 2^{21} paths:

- ▶ **Idea:** Use underlying $P(S_1, S_2)$, whose price $P(0, S_2)$ is known, to correct the estimation of $\text{ZBC}(S_1, S_2, K)$.
- ▶ **Target Payoff (X):**

$$X = e^{-\int_0^{S_1} r_t dt} (P(S_1, S_2) - K)_+$$

- ▶ **Control Variate (Y):**

$$Y = e^{-\int_0^{S_1} r_t dt} P(S_1, S_2), \text{ with known } \mathbb{E}[Y] = P(0, S_2)$$

- ▶ $\text{ZBC}_{CV} = \bar{X} - \beta^*(\bar{Y} - P(0, S_2))$ where $\beta^* = \frac{\widehat{\text{Cov}}(X, Y)}{\widehat{\text{Var}}(Y)}$
- ▶ Variance reduction of 20.4%
- ▶ Simulation time = 1.97 ms
- ▶ Throughput = 1064 M paths/sec

3. Sensitivity Analysis ($\partial_\sigma \mathbf{ZBC}$)

Methodologies

- ▶ **Finite Difference (FD):** Recompute \mathbf{ZBC} with $\sigma \pm \epsilon$.
- ▶ **Pathwise Method (PW):** Differentiate the estimator directly from the analytical formula.

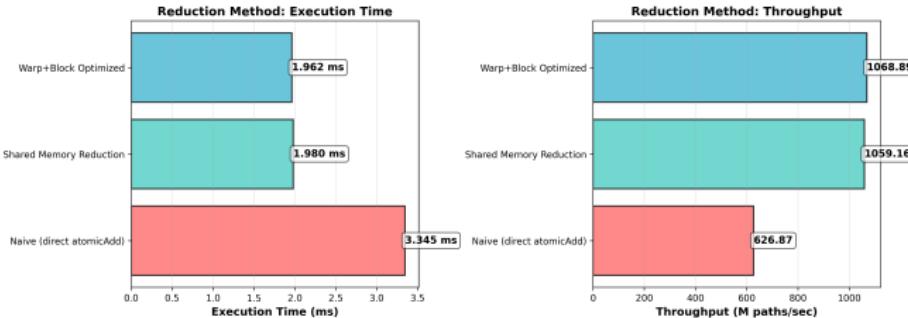
Results

- ▶ **Comparison:** Methods agree within **0.18%**.
- ▶ **PW performance:** 2.06 ms with a throughput of 509 M paths/sec.
- ▶ **Calibration Stability:** In FD need to recalibrate $\theta(t)$, since it depends on σ , and all the values computed from $\theta(t)$, i.e. $m_{s,t}$ and $r(t)$. Instead, a recalibration of $P(0, T)$ leads to an error of 127%

Reduction Benchmark

Comparison of reduction strategies tested for ZBC in question 2:

- ▶ **Naive approach:** every thread directly updates the global sum using atomicAdd;
- ▶ **Shared memory reduction:** Sums accumulated into a shared memory array first. Then, only one thread per block adds the result to global memory.
- ▶ **Warp+block reduction:** First, warp shuffle intrinsics, then via shared memory and finally to global memory.



Thank you for your attention!

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