

Stochastic Deflator for an Economic Scenario Generator with Five Factors

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Outline

- Intro (5 mins)
- Implementation (10 mins)
- Numerical results (5 mins)
- Q&A

Related literature review

- Arrow and Debreu (1954)
- Harrison and Kreps (1979)
- Dastarac and Sauveplane (2010); Caja and Planchet (2011)
- Bonnin et al. (2014); Borel-Mathurin et al. (2015); Vedani et al. (2017)

Motivations

- Many “unusual” scenarios occur (e.g. 10-year rate 50%) under risk-neutral measure, which increases the difficulty to justify the calibration of “reaction functions” embedded in the ALM-projection model used to compute cash flows
- Example

The lapse rate is often a function of the difference between the revalorization rate of the contract and a reference rate; the parameters are calibrated observing “usual” values of economic parameters but may become difficult later to justify for atypical values of economic risk factors

Stochastic Deflator

- Benefit

We could compute best estimate value by simply averaging the multiplication of deflator and projected cash flows

$$E^Q (\delta(t) X) = E^P [D(t) X]$$

$\delta(t)$: discount factor

$D(t)$: stochastic deflator

X : nonnegative random variable

Summary

- In this paper, we implement a stochastic deflator with four economic and financial risk factors: interest rates, stock prices, default intensities, and convenience yields. We examine the deflator with different financial assets, such as stocks, zero-coupon bonds, vanilla options, and corporate coupon bonds.
- Challenges remain for finding better algorithm to approximate the process of deflator.

Brownian Motions

$$\mathbf{W}_{ESG} = \begin{bmatrix} W_r \\ W_S \\ W_\chi \\ W_\gamma \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ \rho_{rS} & \sqrt{1-\rho_{rS}^2} & 0 & 0 \\ \rho_{r\chi} & \rho'_{S\chi} & \rho'_{\chi\chi} & 0 \\ \rho_{r\gamma} & \rho''_{S\gamma} & \rho''_{\chi\gamma} & \rho''_{\gamma\gamma} \end{bmatrix} \begin{bmatrix} W_0 \\ W_1 \\ W_2 \\ W_3 \end{bmatrix}$$

$$\rho'_{S\chi} = \frac{\rho_{S\chi} - \rho_{rS}\rho_{r\chi}}{\sqrt{1-\rho_{rS}^2}}, \quad \rho'_{\chi\chi} = \sqrt{\frac{1-\rho_{rS}^2 - \rho_{r\chi}^2 - \rho_{S\chi}^2 + 2\rho_{rS}\rho_{r\chi}\rho_{S\chi}}{1-\rho_{rS}^2}},$$

$$\rho''_{\chi\gamma} = \frac{\rho_{\chi\gamma} - \rho_{r\chi}\rho_{r\gamma} - \rho_{S\chi}\rho_{S\gamma} - \rho_{rS}^2\rho_{\chi\gamma} + \rho_{rS}\rho_{r\chi}\rho_{S\gamma} + \rho_{rS}\rho_{r\gamma}\rho_{S\chi}}{\sqrt{1+\rho_{rS}^4 - 2\rho_{rS}^3\rho_{r\chi}\rho_{S\chi} - 2\rho_{rS}^2 + \rho_{rS}^2\rho_{r\chi}^2 + \rho_{rS}^2\rho_{S\chi}^2 - \rho_{r\chi}^2 - \rho_{S\chi}^2 + 2\rho_{rS}\rho_{r\chi}\rho_{S\chi}}},$$

$$\rho''_{\gamma\gamma} = \sqrt{1-\rho_{r\gamma}^2 - \rho_{S\gamma}^2 - \rho_{\chi\gamma}^2}.$$

SDEs

$$dr(t) = \alpha(t, r(t))dt + \beta(t, r(t))dW_r(t)$$

$$dB(t) = B(t)r(t)dt$$

$$dP(t, T, r(t)) = P(t, T, r(t)) \left[r(t) + \tilde{\sigma}(t, r(t))\theta(t) \right] dt$$

$$+ P(t, T, r(t)) \tilde{\sigma}(t, r(t)) dW_r(t), \quad \tilde{\sigma}(t, r(t)) = \frac{P_r \beta(t, r(t))}{P(t, r(t))}$$

$r(t)$: interest rate; $B(t)$: short-term saving

$P(t, T, r(t))$: zero coupon bond of no risk with maturity T

$\theta(t)$: market price of risk

SDEs

$$dS(t) = S(t) \mu_S(t) dt + S(t) \sigma_S(t) dW_S(t)$$

$$d\chi(t) = [e - f\chi(t)] dt + \sigma_\chi \sqrt{\chi(t)} dW_\chi(t)$$

$$d\gamma(t) = \eta dW_\gamma(t)$$

$S(t)$: stock price; $\chi(t)$: default density

$\gamma(t)$: convenience yield

SDE of Stochastic Deflator

$$dD(t) = \Omega(D(t), t, r(t))dt + \Phi(D(t), t, r(t))dW_r(t) + \Psi(D(t), t, r(t))dW_1(t) \\ + \Gamma(D(t), t, r(t))dW_2(t) + I(D(t), t, r(t))dW_3(t)$$

$$\Omega(D(t), t, r(t)) = -D(t)r(t), \quad \Phi(D(t), t, r(t)) = -D(t)\theta(t),$$

$$\Psi(D(t), t, r(t)) = \frac{D(t)[r(t) + \theta(t)\sigma_s(t)\rho_{rs} - \mu_s(t)]}{\sigma_s(t)\sqrt{1 - \rho_{rs}^2}},$$

$$\Gamma(D(t), t, r(t)) = D(t) \left[\frac{\theta(t)\rho_{r\chi} + \frac{r(t)\chi(t) - e + f\chi(t)}{\sigma_\chi\rho'_{\chi\chi}\sqrt{\chi(t)}}}{\rho'_{\chi\chi}} + \frac{\rho'_{s\chi}(\mu_s(t) - r(t) - \theta(t)\sigma_s(t)\rho_{rs})}{\rho'_{\chi\chi}\sigma_s(t)\sqrt{1 - \rho_{rs}^2}} \right],$$

$$I(D(t), t, r(t)) = D(t) \left\{ \frac{\frac{\rho_{r\gamma}\theta(t)}{\rho''_{\gamma\gamma}} + \frac{r(t)\gamma(t)}{\eta\rho''_{\gamma\gamma}} - \frac{\rho''_{\chi\gamma}\rho_{r\chi}\theta(t)}{\rho''_{\gamma\gamma}\rho'_{\chi\chi}} + \frac{\rho''_{\chi\gamma}[e - r(t)\chi(t) - f\chi(t)]}{\rho''_{\gamma\gamma}\rho'_{\chi\chi}\sigma_\chi\sqrt{\chi(t)}}}{\rho''_{\gamma\gamma}} + \frac{(\rho''_{s\gamma}\rho'_{\chi\chi} - \rho''_{\chi\gamma}\rho'_{s\chi})[\mu_s(t) - r(t) - \rho_{rs}\theta(t)\sigma_s(t)]}{\rho''_{\gamma\gamma}\rho'_{\chi\chi}\sigma_s(t)\sqrt{(1 - \rho_{rs}^2)}} \right\}$$

Time Discretization

- Euler method, $i = 0, 1, \dots, N - 1$

$$Y_{i+1} = Y_i + b_X(t_i, Y_i)(t_{i+1} - t_i) + \sigma_X(t_i, Y_i)(W_{i+1} - W_i)$$

- Milstein method, $i = 0, 1, \dots, N - 1$

$$Y_{i+1} = Y_i + b_X(t_i, Y_i)(t_{i+1} - t_i) + \sigma_X(t_i, Y_i)(W_{i+1} - W_i) + \frac{1}{2} \sigma_X(t_i, Y_i) \sigma_X(t_i, Y_i) \left[(W_{i+1} - W_i)^2 - (t_{i+1} - t_i) \right]$$

Time Discretization

• Second Milstein method, $i = 0, 1, \dots, N - 1$

$$dX_{t,d \times 1} = a(t, X_t)_{d \times 1} dt + b(t, X_t)_{d \times m} dW_{t,m \times 1}$$

$$df(t, X_t) = \left[\frac{\partial f(t, X_t)}{\partial t} + \sum_{i=1}^d \frac{\partial f(t, X_t)}{\partial x_i} a_i(t, X_t) + \frac{1}{2} \sum_{i,j=1}^d \frac{\partial^2 f(t, X_t)}{\partial x_i \partial x_j} \Sigma_{t,ij} \right] dt + \sum_{i=1}^d \sum_{k=1}^m b_{ik}(t, X_t) \frac{\partial f(t, X_t)}{\partial x_i} dW_{t,k}, \quad \Sigma_t = b(t, X_t) b^T(t, X_t)$$

Time Discretization

•Second Milstein method (continue)

$$L^0 = \frac{\partial}{\partial t} + \sum_{i=1}^d a_i(t, X_t) \frac{\partial}{\partial x_i} + \frac{1}{2} \sum_{i,j=1}^d \Sigma_{t,ij} \frac{\partial^2}{\partial x_i \partial x_j}$$

$$L^k = \sum_{i=1}^d b_{ik}(t, X_t) \frac{\partial}{\partial x_i}, \quad \forall k = 1, \dots, m$$

$$df(t, X_t) = L^0 f(t, X_t) dt + \sum_{k=1}^m L^k f(t, X_t) dW_{t,k}$$

Time Discretization

- Second Milstein method (continue)

$$\begin{aligned} Y_{n+1,i} = & Y_{n,i} + a_i(n, Y_n) \Delta t + \sum_{k=1}^m b_{ik}(n, Y_n) \Delta W_{n,k} + \frac{1}{2} L^0 a_i(n, Y_n) (\Delta t)^2 \\ & + \frac{1}{2} \sum_{k=1}^m \left[L^k a_i(n, Y_n) + L^0 b_{ik}(n, Y_n) \right] \Delta W_{n,k} \Delta t \\ & + \frac{1}{2} \sum_{k=1}^m \sum_{j=1}^m L^j b_{ik}(n, Y_n) (\Delta W_{n,j} \Delta W_{n,k} - V_{jk}) \end{aligned}$$

V_{jk} independent with $\mathbf{Pr}(V_{jk} = \Delta t) = \mathbf{Pr}(V_{jk} = -\Delta t) = \frac{1}{2}$ for $j < k$,

$V_{kj} = -V_{jk}$ for $j > k$, $V_{jk} = \Delta t$ for $j = k$

Numerical results

$$dr(t) = [a_r - b_r r(t)] dt + \sigma_r \sqrt{r(t)} d\tilde{W}_r(t); a_r, b_r, \sigma_r > 0$$
$$d\tilde{W}_r(t) = \theta(t) dt + dW_r(t)$$

Then,

$$dr(t) = [a_r - b_r r(t) + \theta(t) \sigma_r \sqrt{r(t)}] dt + \sigma_r \sqrt{r(t)} dW_r(t)$$
$$d\theta(t) = [a_\theta - b_\theta \theta(t)] dt + \sigma_\theta \sqrt{\theta(t)} dW_\theta(t); a_\theta, b_\theta, \sigma_\theta > 0$$

Numerical results

$$D(0)S(0) = S(0) = E^Q[\delta(T)S(T)] = E^P[D(T)S(T)] = 1$$

$$\begin{aligned} D(0)P(0, T, r(0)) &= P(0, T, r(0)) \\ &= E^Q[\delta(T)P(T, T, r(T))] = E^P[D(T)P(T, T, r(T))] = E^P[D(T)] \end{aligned}$$

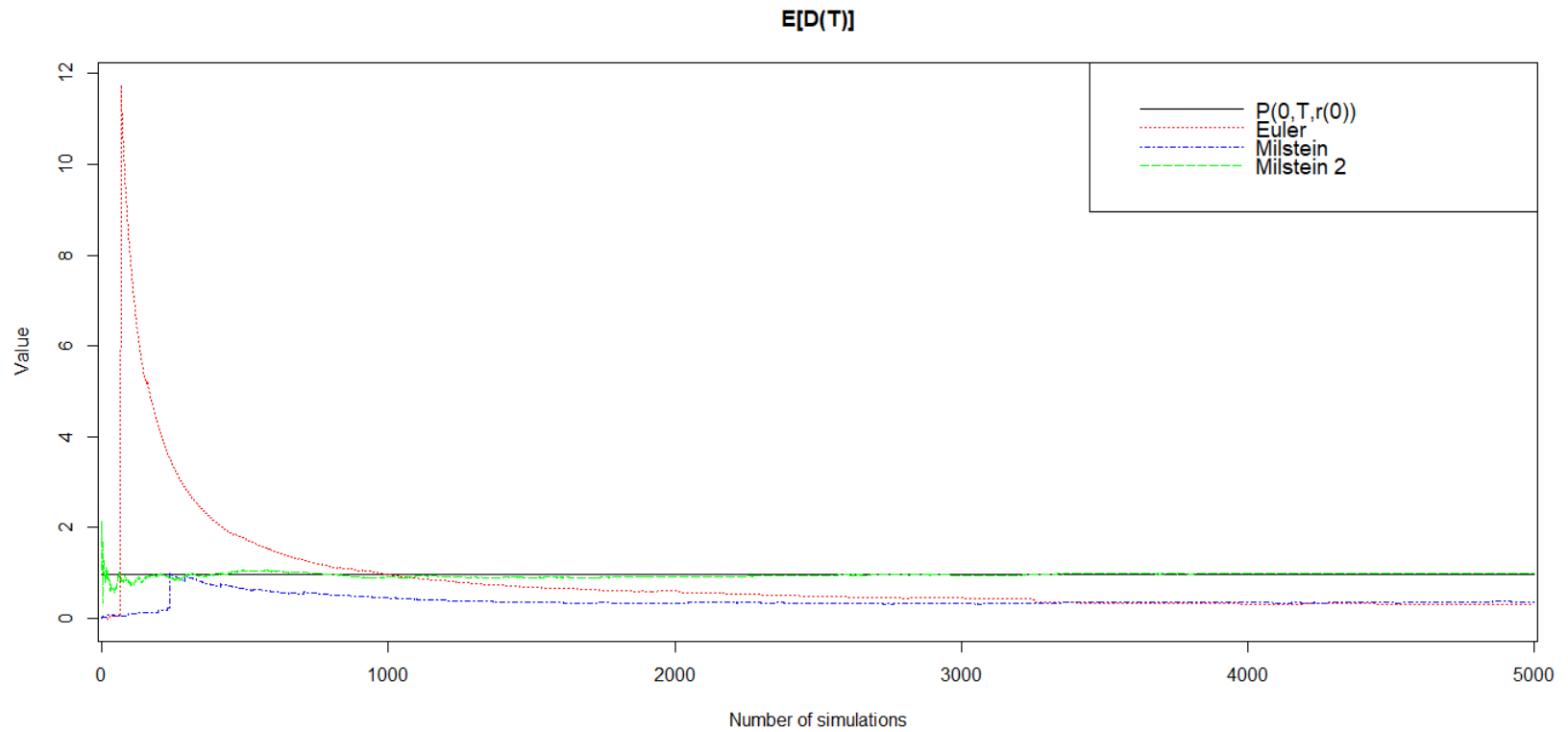
$$\begin{aligned} D(0)Put(r(0), S(0)) &= Put(r(0), S(0)) \\ &= E^Q[\delta(T)(K - S(T))^+] = E^P[D(T)(K - S(T))^+] \end{aligned}$$

Numerical results

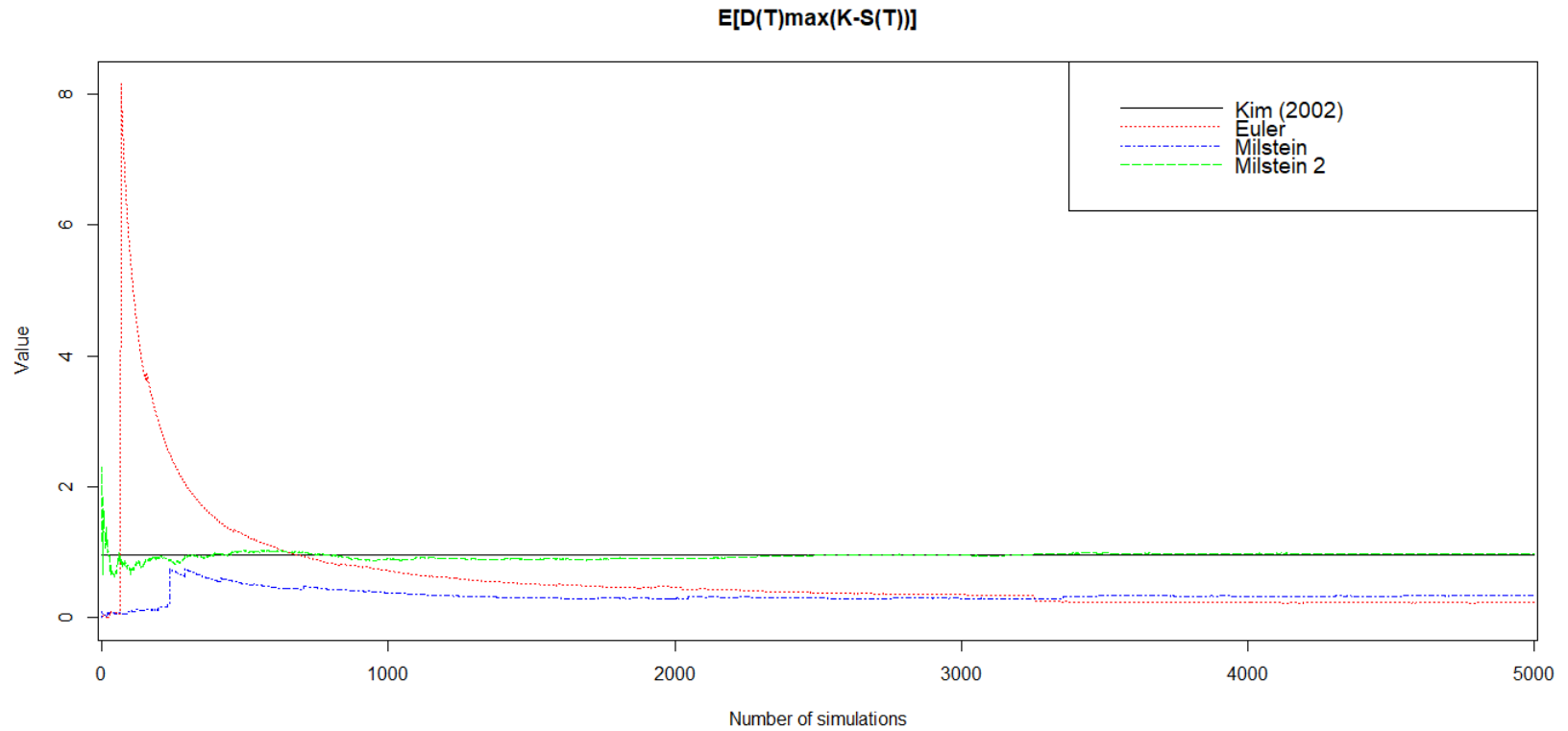
Tab. 2. Zero coupon bond of no risk with maturity T ↵

# of Simulations ↵	$E[D(T)]$ ↵	$E[D(T)P(T, T, r(T))]$ ↵	$P(0, T, r(0))$ ↵
Euler method ↵			
2500 ↵	0.48812434395029 ↵	0.488066560145415 ↵	↵
5000 ↵	0.308038513966897 ↵	0.308003081851407 ↵	↵
10000 ↵	0.325024763106268 ↵	0.324990036772998 ↵	↵
100000 ↵	0.625985956594468 ↵	0.625922227828396 ↵	0.970957220487724 ↵
250000 ↵	0.771419872927604 ↵	0.771346645830578 ↵	
500000 ↵	0.726182177029966 ↵	0.72611342647583 ↵	
1000000 ↵	1.18961111433854 ↵	1.18949806538064 ↵	
Milstein method ↵			
2500 ↵	0.332402071677238 ↵	0.332366375930865 ↵	↵
5000 ↵	0.366969978004539 ↵	0.366931013968048 ↵	↵
10000 ↵	0.360964352720743 ↵	0.360926322954742 ↵	↵
100000 ↵	0.654420842856107 ↵	0.654354469594408 ↵	0.970957220487724 ↵
250000 ↵	0.746857198318386 ↵	0.746783090935181 ↵	
500000 ↵	0.824192927498608 ↵	0.82411348941373 ↵	
1000000 ↵	1.15968204198937 ↵	1.15956727600934 ↵	
Second Milstein method ↵			
2500 ↵	0.952656631120277 ↵	0.952657151589584 ↵	↵
5000 ↵	0.986807314190818 ↵	0.986807555270611 ↵	↵
10000 ↵	1.00104803675261 ↵	1.00104821737114 ↵	↵
100000 ↵	0.97943249260489 ↵	0.979432483928667 ↵	0.970957220487724 ↵
250000 ↵	0.974542088257683 ↵	0.974542020575479 ↵	
500000 ↵	0.975959099152459 ↵	0.975959075808478 ↵	
1000000 ↵	0.973919234689554 ↵	0.97391925959157 ↵	

Numerical results



Numerical results



Numerical results

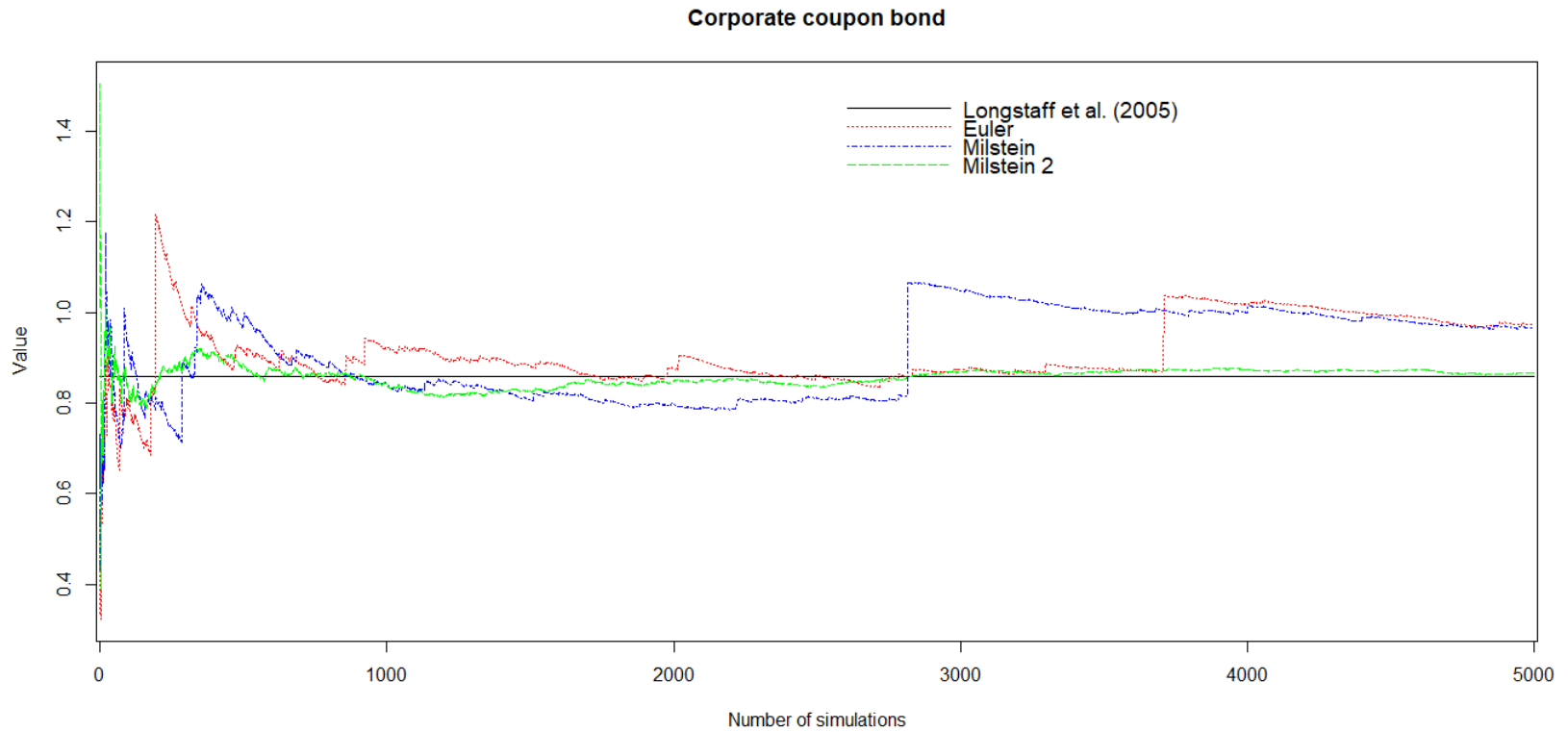
Longstaff et al. (2005)

$$\rho_{r\chi} = 0, \rho_{r\gamma} = 0, \rho_{S\chi} = 0, \rho_{S\gamma} = 0, \rho_{\chi\gamma} = 0$$

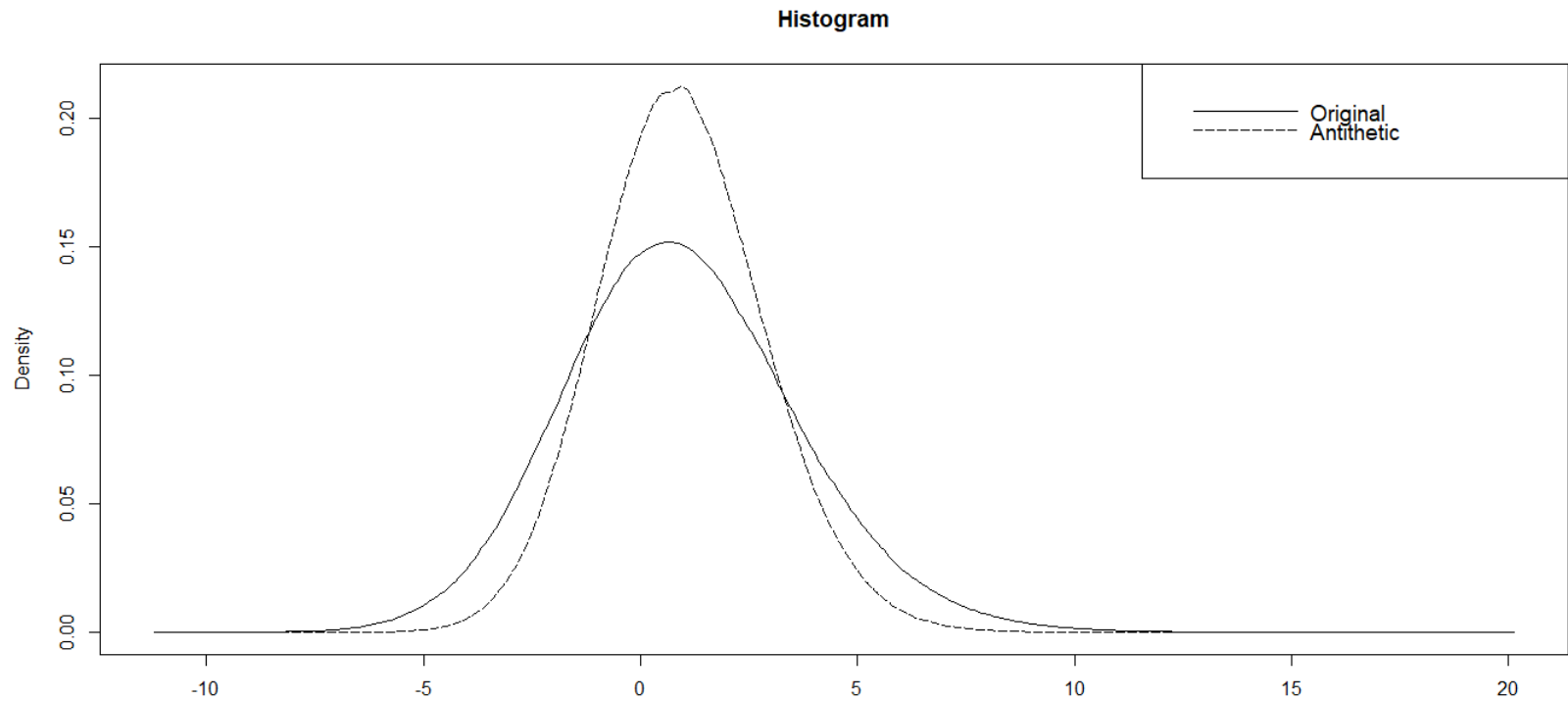
$$dB(t) = B(t)[r(t) + \chi(t) + \gamma(t)]dt$$

$$D(0)CB(c, \omega, T) = E^{\mathbf{P}} [D(T)] + cE^{\mathbf{P}} \left[\int_0^T D(t) dt \right] + (1 - \omega)E^{\mathbf{P}} \left[\int_0^T \chi_t D(t) dt \right]$$

Numerical results



Antithetic Sampling



Confidence Interval

$$CI_{95\%} = E^P [D(t)] + \frac{\text{Var}(E^P [D(t)])}{2} \pm t_{d.f.=n} \sqrt{\frac{\text{Var}(E^P [D(t)])}{n} + \frac{[\text{Var}(E^P [D(t)])]^2}{2(n-1)}}$$

Second Milstein method with antithetic sampling and sample size equalling 2500

$$CI_{95\%} \text{ of } E^P [D(T)] = [0.95192, 0.95559]$$

Q&A

Thank you!