

A discount curve for Insurance Risk Management with exact fit and parsimonious forecasts

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Plan

1. Context
2. Curve construction and extrapolation
 - ▶ Curve construction
 - ▶ Curve extrapolation
3. Real world forecasting

1. Context

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2. Curve construction and extrapolation
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3. Real world forecasting

1. Context

- ▶ Calculation of insurers' technical provisions
- ▶ Discount curve **calibration** and **extrapolation** for pricing and risk neutral simulation
- ▶ Curves' **real world forecasting** for risk management

2. Curve construction and extrapolation

2.1. Curve construction

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2. Curve construction and extrapolation

2.1. Curve construction

- ▶ Discount curve construction: uses IRS + Credit Risk Adjustment (Solvency II framework) or OIS
- ▶ Curve construction methodology: ideas from Schlögl and Schlögl (2000)
- ▶ We add:
 - ▶ static discount curve **extrapolation**
 - ▶ curves' **real world forecasting**
- ▶ Can be extended to multiple curve construction

2. Curve construction and extrapolation

2.1. Curve construction (cont'd)

- ▶ **Idea:** No arbitrage short rate models include $t \mapsto b(t)$ for exact **fitting of initial term structure**
- ▶ **Examples:**
 - ▶ Hull & White (1990, 1994) extended Vasicek
 - ▶ Extended CIR...
- ▶ **piecewise-constant specification** of $t \mapsto b(t)$ at input swaps' maturities $T_1, \dots, T_n \longrightarrow b_1, \dots, b_n$

2. Curve construction and extrapolation

2.1. Curve construction (cont'd)

- ▶ **Discount factors** in the extended Vasicek case:

$$P(0, t) = \exp \left(-X_0 \phi(t) - a \int_{t_0}^t b(u) \phi(t-u) du - \psi(t) \right)$$

Where:

$$\phi(s) := \frac{1}{a} (1 - e^{-as})$$

$$\psi(s) := - \int_0^s \left(\frac{\sigma^2}{2} \phi^2(s-\theta) \right) d\theta$$

- ▶ **Piecewise-constant** $t \mapsto b(t)$:

$$\int_0^t b(u) \phi(t-u) du \longrightarrow \sum_{k=1}^n b_k \int_{T_{k-1} \wedge t}^{T_k \wedge t} \phi(t-u) du \longrightarrow P^M(0, t)$$

2. Curve construction and extrapolation

2.1. Curve construction (cont'd)

- ▶ Calibration?
 - ▶ S = swaps values at time $t = 0$
 - ▶ C = swaps cash flows matrix
 - ▶ W = diagonal matrix of minimization weights
 - ▶ \mathbf{P} = **unknown** discount factors, function of b_1, \dots, b_n
- ▶ From **less bias (and more variance)** to **more bias (and less variance)**
 - ▶ Solve $S = \mathbf{CP}$ iteratively (a.k.a *bootstrapping*)
 - ▶ Minimize $(S - \mathbf{CP})^T W (S - \mathbf{CP})$
 - ▶ Minimize $(S - \mathbf{CP})^T W (S - \mathbf{CP}) + \text{Penalization}$

2. Curve construction and extrapolation

2.2. Curve extrapolation

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2. Curve construction and extrapolation

2.2. Curve extrapolation

- ▶ In the Extended Vasicek case, for $t > T_n$ add a parameter:

$$b_{n+1} = f_{\infty} + \frac{\sigma^2}{2a^2}$$

where $f_{\infty} = UFR = \textit{Ultimate Forward Rate}$

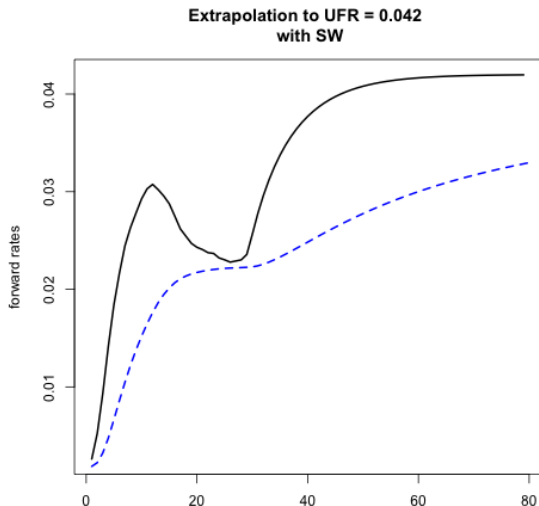
- ▶ Either **exogenously fixed UFR** and convergence period $\tau_{convergence}$: increase parameter a until

$$|f^M(0, LLP + \tau_{convergence}) - f_{\infty}| < \textit{tolerance}$$

- ▶ Or **data driven UFR**:
 - ▶ A fraction of the swaps data in a **training set**, the other in a **test set**
 - ▶ A grid search on values for f_{∞} \longrightarrow lowest pricing error on swaps from the **test set**

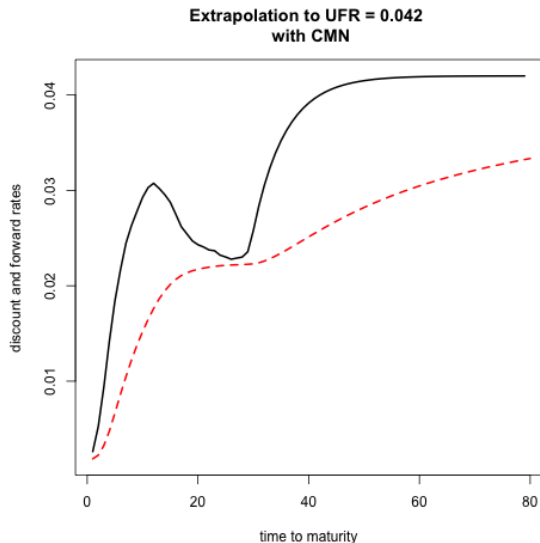
2. Curve construction and extrapolation

- ▶ IRS from Ametrano and Bianchetti (2013) + CRA = 10bps
- ▶ **Fixed UFR = 4.2%**, $LLP = 20$ years, convergence period $T_{convergence} = 40$ years, **Smith-Wilson method**:



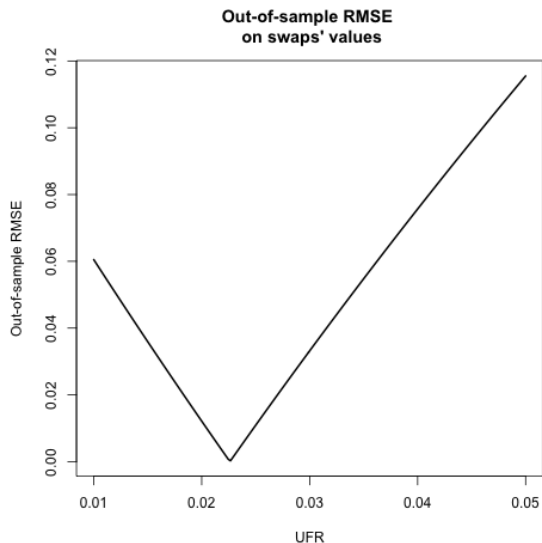
2. Curve construction and extrapolation

- ▶ **Fixed UFR = 4.2%**, $LLP = 20$ years, convergence period $\tau_{convergence} = 40$ years, the **method presented here**:



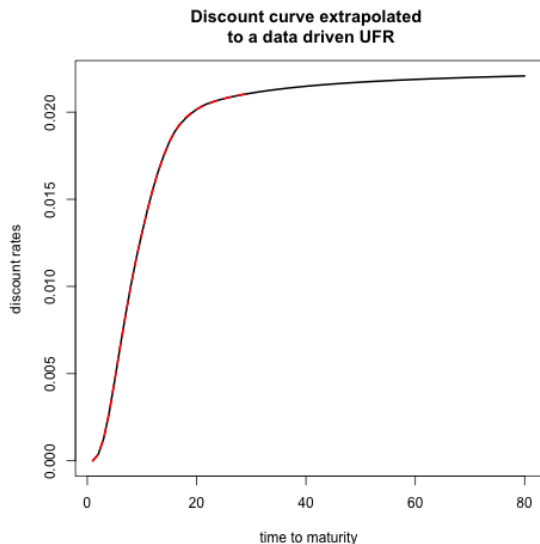
2. Curve construction and extrapolation

- ▶ OIS from Ametrano and Bianchetti (2013), **data driven UFR** = 2.26%:



2. Curve construction and extrapolation

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3. Real world forecasting

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3. Real world forecasting

- ▶ **Problem:** Forecasting the constructed 'discretized HW' curves to future dates, in real world probability
- ▶ **A solution:** problem's **dimension reduction**
 - ▶ Functional Principal Components Analysis (**FPCA**) (Ramsay and Silverman (2005)) on parameters
 - ▶ Calibrate spot rates through time; obtain:

$$\mathbf{B}_{ij} = \hat{b}_{x_i}(t_j)$$

- ▶ Based on the *cross-product function*:

$$\mathbf{V} = \frac{1}{N} \mathbf{B}^T \mathbf{B}$$

find Functional PCs: ξ_1, \dots, ξ_K

3. Real world forecasting (cont'd)

▶ **A solution (cont'd):**

- ▶ Express **calibrated** $\hat{b}_{x_i}(t_j)$ as:

$$\hat{b}_{x_i}(t_j) = \beta_{t_j,0} + \sum_{k=1}^K \beta_{t_j,k} \xi_k(x_i) + \epsilon_{x_i}(t_j)$$

- ▶ Obtain forecasts for $\hat{b}_{x_i}(t_j)$ at $t > t_j$: **univariate time series forecasts** on $\beta_{t_j,k}$

$$\beta_{t_j,k} \longrightarrow \hat{\beta}_{t_j+h|t_j,k}$$

3. Real world forecasting (cont'd)

- ▶ **A solution (cont'd):**

- ▶ Real world simulation: parametric law or **bootstrap resampling (with replacement)** of univariate time series **residuals**

- ▶ Bootstrap:

$$\hat{\beta}_{t_j, k} \longrightarrow \hat{\beta}_{t_j, k}^*$$

- ▶ Obtain univariate forecasts:

$$\hat{\beta}_{t_j + h | t_j, k}^*$$

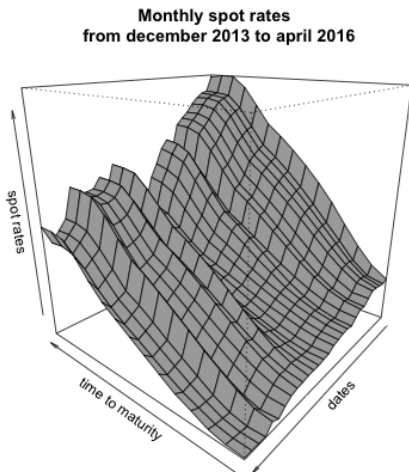
- ▶ And:

$$\hat{b}_{x_i}^*(t_j + h)$$

- ▶ Plug $\hat{b}_{x_i}^*(t_j + h)$ into formulas for discount factors

3. Real world forecasting (cont'd)

- ▶ IRS + CRA data. **12-months ahead forecasts**, **1000 simulations**, **FPCA** and **bootstrap resampling**:



3. Real world forecasting (cont'd)

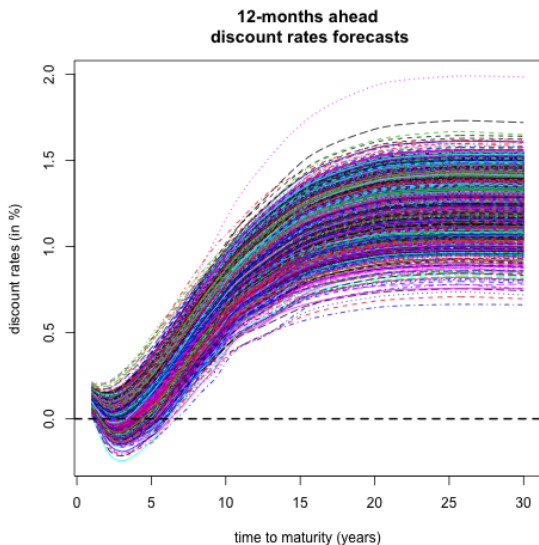
- ▶ Benchmarking with models *à la* Diebold-Li (2006)
- ▶ Reasonable forecasts. But benchmarks are **subjective**
- ▶ In practice: study the univariate time series

Model	Avg. OOS error
CMN - auto.arima	0.0031
CMN - ets	0.0037
NS - auto.arima	0.0031
NS - ets	0.0035
NSS - auto.arima	0.0027
NSS - ets	0.0035

- ▶ Next: use of AR(1)

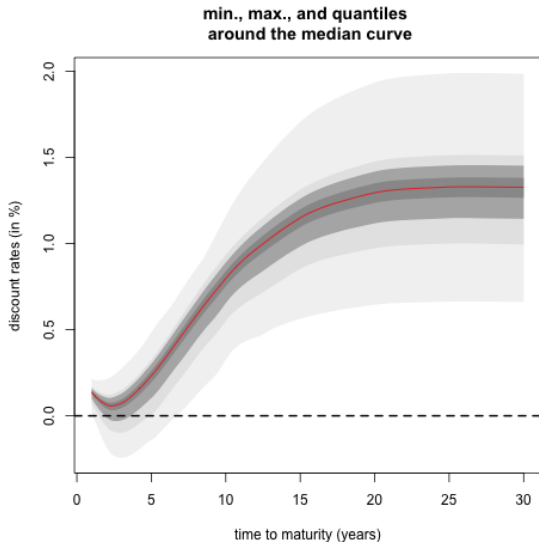
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3. Real world forecasting (cont'd)

- ▶ IRS + CRA data. **12-months ahead forecasts, 1000 simulations, FPCA** and **bootstrap resampling**:



Questions

- ▶ Questions?

References

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