

# **Chapter 2: Estimating the Term Structure**

## **2.3 Smoothing Methods**

### **Interest Rate Models**

Damir Filipović

## 2.3 Smoothing Methods

Exact methods:

- Discount curve is sensitive to small changes in cash flow matrix.
- Discount factors of similar maturity can be very different.
- Leads to ragged forward curves.

Smoothing methods: estimate a smooth forward curve from market rates at the cost of not exactly matching data.

Set spot date  $t_0 = 0$  for simplicity.

Data:  $N$  yields  $y_i = y(0, T_i)$  with maturities  $0 < T_1 < \dots < T_N$ .

Aim: find smooth forward curve  $f(T) = f(0, T)$  matching the yields optimally

$$\int_0^{T_i} f(u) du = T_i y_i + \epsilon_i$$

for pricing errors  $\epsilon_i$  subject to being minimized.

Smoothing methods mainly used by central banks.

- Smoothness: supply a market expectation for monetary policy purposes rather than precise pricing of all bonds in the market.
- Flexibility: sufficiently flexible to capture movements in the underlying term structure.
- Stability: small changes in data at one maturity do not have disproportionate effect on forward rates at other maturities.

# Estimation Methods Used by Several Central Banks

Bank for International Settlements 2005: Nelson–Siegel (NS), Svensson (S), weighted prices (wp).

Central bank	Method	Minimized error
Belgium	S or NS	wp
Canada	Exponential spline	wp
Finland	NS	wp
France	S or NS	wp
Germany	S	yields
Italy	NS	wp
Japan	Smoothing spline	prices
Norway	S	yields
Spain	S	wp
Sweden	Smoothing spline or S	yields
Switzerland	S	yields
UK	Smoothing spline	yields
USA	Smoothing spline	bills: wp, bonds: prices

Parametric families of forward curves.

- Nelson–Siegel:

$$f(T) = \beta_0 + \beta_1 e^{-aT} + \beta_2 a T e^{-aT}$$

for parameters  $\beta_0, \beta_1, \beta_2$ , and  $a$ .

- Svensson:

$$f(T) = \beta_0 + \beta_1 e^{-a_1 T} + \beta_2 a_1 T e^{-a_1 T} + \beta_3 a_2 T e^{-a_2 T}$$

for parameters  $\beta_0, \beta_1, \beta_2, \beta_3$ , and  $a_1, a_2$ .

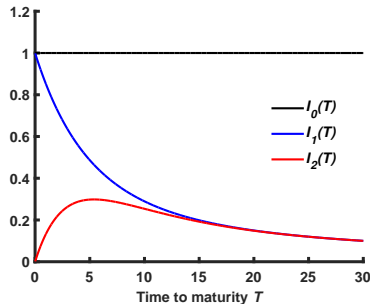
Nelson–Siegel yield curves are

$$\begin{aligned}y(T) &= \frac{1}{T} \int_0^T f(u) du \\ &= \beta_0 l_0(T) + \beta_1 l_1(T) + \beta_2 l_2(T)\end{aligned}$$

with basis functions

- $l_0(T) = 1$  (level)
- $l_1(T) = \frac{1-e^{-aT}}{aT}$  (slope)
- $l_2(T) = \frac{1-e^{-aT}}{aT} - e^{-aT}$  (curvature).

Figure with  $a = 0.3$ :



Issues: flexibility, stability

Find forward curve in Hilbert space  $H$  consisting of absolutely continuous functions on  $[0, T_*]$  with scalar product

$$\langle g, h \rangle_H = g(0)h(0) + \int_0^{T_*} g'(u)h'(u) du.$$

Solve optimization problem

$$\min_{f \in H} \int_0^{T_*} (f'(u))^2 du + \alpha \sum_{i=1}^N \left( T_i y_i - \int_0^{T_i} f(u) du \right)^2 \quad (P)$$

where  $\alpha > 0$  tunes trade-off between smoothness and correctness of fit.



The unique solution  $f$  of (P) is a quadratic spline characterized by

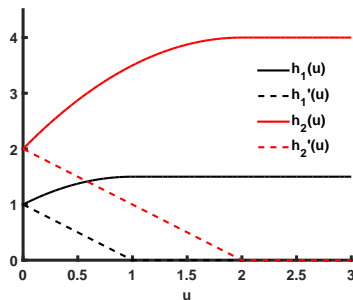
$$f(u) = \beta_0 + \sum_{i=1}^N \beta_i h_i(u)$$

where  $h_i \in C^1[0, T_*]$  is a quadratic basis spline with

$$h'_i(u) = (T_i - u)^+, \quad h_i(0) = T_i$$

and  $\beta_0, \dots, \beta_N$  solve a linear system.

Quadratic basis splines  $h_1$  and  $h_2$  together with first derivatives for  $T_1 = 1$  and  $T_2 = 2$ :



Lorimier's theorem continued:  $\beta_0, \dots, \beta_N$  solve the linear system of equations

$$\sum_{i=1}^N \beta_i T_i = 0,$$
$$\alpha \left( y_i T_i - \beta_0 T_i - \sum_{j=1}^N \beta_j \langle h_i, h_j \rangle_H \right) = \beta_i, \quad i = 1, \dots, N.$$

Smoothing splines satisfy key criteria: smoothness, flexibility, stability.

Parameter  $\alpha$  tunes trade-off between smoothness and correctness of fit:

- $\alpha \rightarrow 0$ : maximal smoothness, constant forward curve  $f(T) = \beta_0$
- $\alpha \rightarrow \infty$ : perfect fit subject to minimal  $\int_0^{T^*} (f'(u))^2 du$

Choice of  $\alpha$  is critical.

Example: Swiss government bond yields in August 2011

