

Discussion of Score-Driven Nelson Siegel: Hedging Long-Term Liabilities

Michel Vellekoop

University of Amsterdam and Netspar



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Nelson-Siegel Dynamics

Dynamic model for $p(t, T)$, the zero-coupon bond prices at time t for maturity $T \geq t$:

$$p(t, t + \tau) = \exp(-\tau B_{\lambda_t}(\tau) \tilde{f}_t) = \exp(-b_{\lambda_t}(\tau) \tilde{f}_t)$$

which are driven by a three-dimensional (column) vector process \tilde{f}_t .



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The Nelson Siegel functions

$$b_{\lambda}(\tau) = \left[\tau, \frac{1}{\lambda}(1 - e^{-\lambda\tau}), \frac{1}{\lambda}(1 - e^{-\lambda\tau}) - \tau e^{-\lambda\tau} \right].$$

depend on a shape parameter λ which in the model is allowed to vary over time: λ_t .



Dynamic Conditional Score Model

Goal is hedging: i.e. reduced sensitivity to interest rate shocks for portfolio which contains bonds and long-term liabilities.

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$$\lambda_t - \phi_0 = \phi_1(\lambda_{t-1} - \phi_0) + u_{t-1},$$

with u_{t-1} a deterministic function of λ_{t-1} and other observed quantities at time $t - 1$ (scaled score function of observations wrt λ_t).

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- Idiosyncratic risks modeled by single GARCH process with equal weights for all maturities.



Results

- Cases considered
 - maturities up till 20 years or full set (i.e. also 25, 30, 40 and 50 years maturity)
 - constant or GARCH process for idiosyncratic volatility
 - constant or DCS process for shape parameter
- Key result: hedging 50-year bond using bonds with maturities of 20 or less

Table: Reduction in Variance of Portfolio Returns

88.2%	First factor only
92.6%	NS factors, shape fixed, GARCH vol
92.7%	NS factors, shape fixed, constant vol
93.8%	NS factors, shape time-varying, constant vol
93.9%	NS factors, shape time-varying, GARCH vol



Arbitrage

- Unadjusted form of Nelson-Siegel model with time-varying parameters, so there may be arbitrage opportunities.
- Not clear how the [Christensen et al. \(2011\)](#) correction term which avoids this would impact the hedging performance, since
 - that term strongly depends on the shape parameter which is no longer taken constant, and
 - examples in that paper suggest that adjustments are highest for largest maturities.



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 - No information criterion values (AIC/BIC) reported.
- May be interesting to compare hedging efficiency with other approaches that vary shape parameter:
 - Generalized Autoregressive Score approach of [Creal et al. \(2008\)](#) for factor model,
 - Extended Kalman Filter of [Koopman et al. \(2010\)](#) for piecewise constant/cubic spline model, or
 - Approximate Filter Algorithm of [Hevia et al. \(2015\)](#) for regime switching model.

or use multivariate time-varying volatility structures and heavy tails, as in [Koopman, Lucas and Zamojski \(2016\)](#).



Application: hedging long-term pension liabilities

- Portfolio to hedge 50-year maturity exposures involves very large positions:
 - Median position for 20-year bond is more than 150%, and
 - 95% quantiles of -400% around maturity 7 and $+500\%$ at maturity 20.



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- Perhaps more relevant to hedge movements in mid-range maturities?

