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Forecasting the term structure of interest rates in Slovakia

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- * **Q**: Can we predict yields on Slovak government bonds better than a naive benchmark out-of-sample?
- ★ Why is this an important question?
 - 1. Assess the role of active government debt management
 - 2. Slovak bonds are different from other major government bond markets (U.S., U.K.) due to credit and liquidity premiums, and lack of domestic monetary policy tools
- Focus on a small number of forecasting models and longer horizons (e.g. one to three years)

This paper

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- \star We split the forecasting exercise into two parts:
 - 1. Forecasting of German government bond yields
 - 2. Forecasting of the spread of Slovak to German government bonds
- * The sum-of-the-parts forecasting model delivers better forecasting performance than the random walk, especially at long horizons
- * Source of improvement is the negative correlation between spread forecast errors and forecast errors on German government bonds

Related literature

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Expectations hypothesis. Fama & Bliss (1987), Campbell & Shiller (1991), [regressions on yield curve slope]; Cieslak & Povala (2015) [yield curve decomposition into short rate expectations and term premium];

Statistical models Diebold & Li (2006) [dynamic Nelson-Siegel]; Ang & Piazzesi (2003) [add macro factors to a reduced-form no-arbitrage term structure model]; de Pooter & Ravazzolo & van Dick (2010) [adding macro factors improves forecasting performance in recessions];

Forecasting credit spreads. Amstad & Remolona & Shek (2016), Cornelli (2012) [relative importance of country fundamentals vs. global financial variables, e.g. VIX]; Ejsing & Grothe & Grothe (2015) [credit and liquidity spreads during the eurozone sovereign debt crisis].



- Sample period January 2003–December 2016
- Estimated zero-coupon yields on Slovak government bonds available on the National Bank of Slovakia website
- Number of regime changes (i)euro adoption, (ii) two financial crises, and (iii) non-conventional monetary policy \longrightarrow complicates forecasting



- Credit spread is an important driver of yields \longrightarrow in the post-2011 period the main source of variation in yields
- Eurozone crisis in 2011-2012 more severe than the Great Recesssion in 2008

Currently low spreads not unique

*Credit spread before 2009 is obtained from international EUR-denominated bonds.

Framework

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- Term structure of interest rates: market expectations about future policy rates + risk premiums
- * Key components of the Slovak yield curve:
 - 1. Expectations about the ECB policy rate
 - 2. Term premium
 - 3. Sovereign credit risk premium
 - 4. Liquidity risk premium
- Need a forecasting model that takes the economic decomposition of the yield curve into account

Models

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$$y_t^{(n)} = \underbrace{\frac{1}{n} \sum_{i=0}^{n-1} E_t r_{t+i}}_{i=0} + \underbrace{\frac{1}{n} \sum_{i=1}^{n} E_t \pi_{t+1}}_{\text{term premium}} + \underbrace{\frac{t p_t^{(n)}}_{t\text{term premium}}}_{\text{term premium}}$$

expectation hypothesis term

 Approximate inflation expectations with a discounted moving average of past inflation:

$$\tau_t^{CPI} = (1 - \nu) \sum_{i=0}^{t-1} \nu^i \pi_{t-i}$$

- * Use one-period yield to extract the real rate (no term premium)
- Term premium is a residual from regressing long-term yield on inflation expectations and the real rate

Models (cont'd)

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- ★ Forecasting the three components:
 - 1. Inflation expectations τ_t^{CPI} : assume random walk
 - 2. Real rate: assume AR(1) process
 - 3. Term premium: assume AR(1) process
- □ Compare the forecast to the dynamic Nelson-Siegel model widely used in the literature

□ Consider three models for forecasting spreads:

- 1. Random walk
- 2. AR(1) model
- 3. "Extended" model: lagged spread, slope of German curve, VIX

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- * Forecasting horizons: one, six, 12, 24, and 36 months
- * Benchmark forecast: random walk $\hat{y}_{t+h|t,rw}^{(\tau)} = y_t^{(\tau)}$
- ★ Metric: ratio of Root Mean Squared Forecast Errors (RMSFE) to random walk RMSFE
- * Use Diebold-Mariano test to evaluate the significance
- * Out-of-sample period: January 2009 through December 2016

DNS German yields						
	1-month	6-month	12-month	24-month	36-month	
1Y	1.27*	1.41	1.79	2.59	3.19	
2Y	1.05	1.23	1.61	2.23	2.47	
3Y	1.02	1.25	1.57*	2.03	2.13*	
4Y	1.11*	1.30*	1.57*	1.90	1.94*	
5Y	1.22*	1.32*	1.54*	1.79*	1.80*	
6Y	1.23**	1.31*	1.49*	1.69*	1.69*	
7Y	1.16*	1.26*	1.43*	1.59*	1.60*	
8Y	1.06*	1.20*	1.35*	1.49*	1.51*	
9Y	1.00	1.13*	1.26*	1.40*	1.43*	
10Y	1.01	1.06	1.18	1.32	1.36*	

* indicates that the difference is statistically significant as indicated by Diebold-Mariano test.

- ★ DNS model significantly under-performs the naive benchmark
- * Deterioration in performance mainly due to zero lower bound

Restr. cycles (German bonds)					
	1-month	6-month	12-month	24-month	36-month
1Y	0.99	0.99	1.11	1.98*	2.37*
2Y	0.99	1.02	1.14	1.66*	1.73*
3Y	1.00	1.03	1.11	1.41	1.37
4Y	1.00	1.03	1.07	1.22	1.15
5Y	1.00	1.02	1.04	1.09	1.01
6Y	1.00	1.01	1.00	1.00	0.91
7Y	1.00	1.00	0.98	0.93	0.83*
8Y	1.00	0.99	0.95	0.87	0.78*
9Y	1.00	0.99	0.94	0.83*	0.73*
10Y	1.00	0.98	0.92	0.79*	0.70*

★ Modeling components of the yield curve improves forecasting performance

* Best model restricts term premium component to be zero, i.e. forecast real rate

	Restr. cycles (German bonds) + RW (spreads)					
	1-month	6-month	12-month	24-month	36-month	
1Y	0.66	0.70	0.86	1.18*	1.58*	
2Y	0.69	0.74	0.89	1.13*	1.25*	
3Y	0.70	0.75	0.87	1.02*	1.07*	
4Y	0.71	0.74	0.84	0.92*	0.95	
5Y	0.71	0.74	0.81	0.84	0.87	
6Y	0.72	0.74	0.79	0.79	0.81	
7Y	0.73	0.74	0.77	0.74	0.77*	
8Y	0.73	0.74	0.76	0.71	0.73*	
9Y	0.74	0.74	0.76	0.69	0.71*	
10Y	0.75	0.75	0.75	0.67	0.69*	

- \star Any of forecasting model of spreads produces significantly lower forecasting errors than a naive forecast
- ★ Interaction of credit spread and German government bond yield forecast errors improves the forecasting performance

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- * Yields on Slovak government bonds are predictable out-of-sample
- The key to this result is an economically-motivated decomposition of yields
- * Source of predictability is the negative correlation of forecast errors

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Data

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Data, Robustness

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Data

- □ Zero-coupon yield curve extracted from Slovak government bonds, monthly frequency, 2003-2016, from the National Bank of Slovakia
- □ German zero-coupon yield curve, monthly frequency, 1991-2016, from the Bundesbank