# Can Affine Term Structure Models Help Us to Predict Exchange Rates? - Discussion

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## Outline

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Objectives Framework Results

### Forward Premium Puzzle

Uncovered Interest Rate Parity:

$$\Delta s_{t+1} = \alpha_0 + \alpha_1 \left[ \mathbf{r}_t - \mathbf{r}_t^{\star} \right] + \varepsilon_{t+1},$$

where  $s_t$  is in  $\frac{1}{\sqrt{1}}$  where  $s_t$  is in  $\frac{1}{\sqrt{1}}$ 

- UIP condition:  $\alpha_0 = 0$  and  $\alpha_1 = 1$ .
- In the data:  $\alpha_1 < 1$  and mostly  $\alpha_1 < 0$ .

Objectives Framework Results

## Exchange Rate Predictability

- Exchange rates are hard to predict.
- Meese and Rogoff (1983): random walk beats macro models for out-of-sample forecasts at short horizons.
  - Clarida and Taylor (1997): VECM beats random walk.
- ► Longer horizons: Mark (1995).
  - Kilian (1999): results not robust.

Summary Objectives Comments Framework Conclusion Results

#### This Paper: An Internationally Affine Model

• State vector: 
$$X_t = \begin{bmatrix} r_t \\ r_t^* \end{bmatrix}$$
.

- ► X<sub>t</sub> follows an Orstein-Uhlenbeck process.
- Market price of risk  $\Lambda_t$  affine in  $X_t$ :

$$dX_t = \Phi(\Theta - X_t) + \Sigma^{1/2} dW_t$$
  

$$\Lambda_t = \Lambda_0 + \Lambda_1 X_t$$
  

$$\Lambda_t^* = \Lambda_0^* + \Lambda_1^* X_t.$$

SummaryObjectivesCommentsFrameworkConclusionResults

## Real Exchange Rate in Complete Markets

For a foreign investor buying a bond from her country, the real return R<sup>\*</sup><sub>t,t+1</sub> satisfies:

$$E_t(M_{t,t+1}^{\star}R_{t,t+1}^{*}) = 1.$$

But a domestic investor can also buy a foreign bond:

$$E_t(M_{t,t+1}\frac{Q_{t+1}}{Q_t}R_{t,t+1}^{\star})=1.$$

▶ Thus, in complete markets, real exchange rate *Q* is defined as:

$$\frac{Q_{t+1}}{Q_t} = \frac{M_{t,t+1}^\star}{M_{t,t+1}}$$

Objectives Framework Results

# Procedure

No arbitrage in bond and FX markets.

Interest rates:

$$r_t^h = A(h) + B(h)'X_t$$
  

$$r_t^{\star,h} = A^{\star}(h) + B^{\star}(h)'X_t, \text{ where } X_t = \begin{bmatrix} r_t \\ r_t^{\star} \end{bmatrix}.$$

Exchange rates:

$$\Delta s_{t+1} = C(1) + D(1)'\widetilde{X}_t + v_{t+1}, ext{ where } \widetilde{X}_t = [X'_t, extsf{vech}(X_tX'_t)']'.$$

#### MLE.

Objectives Framework Results

#### Results

- Forward bias:
  - US-Canada, 3-month horizon:
    - $\alpha_1 = -0.5$  in the model = -0.8 in the data.

US-UK, 3-month horizon:

- $\alpha_1 = -1.9$  in the model = -1.5 in the data.
- Exchange rate predictability out-of-sample:
  - US-Canada, 9% lower RMSE (vs random walk and VAR) at 12-month
  - ▶ US-UK: 36% lower.

Interpretation Questions Suggestions

## Two possible mechanisms

- Lustig-Verdelhan (2005).
- Complete markets, no inflation risk.
- Log currency risk premium:

$$std_{t}m_{t+1}\left[std_{t}m_{t+1}-\rho_{t}\left(m_{t+1},m_{t+1}^{\star}\right)std_{t}m_{t+1}^{\star}\right]$$

Two possible mechanisms: lower foreign interest rate means

- Heteroskedasticity:  $std_t m_{t+1}^* \nearrow$
- Time-varying correlation:  $\rho_t(m_{t+1}, m_{t+1}^{\star}) \nearrow$
- ▶ ⇒ Both mechanisms are playing here (see signs of estimated  $\lambda_{i,j}$ )

Interpretation Questions Suggestions

# US-UK

• 
$$M_{t,t+1} = e^{-r_t} e^{-\lambda_t X_{t+1}} / \phi^P(-\lambda_t, X_t).$$

• 
$$std_t m_{t+1}$$
 is prop. to  $-\lambda_t$ .

- As a first approximation, take  $dr_t \perp dr_t^{\star}$ .
- Estimation (Table 3 in the paper):

$$\Lambda_t = 5.7 - 2.1r_t - 0.6r_t^* \Lambda_t^* = -5.9 + 2.6r_t + 0.3r_t^*$$

$$\ \, r_t^\star \searrow \Longrightarrow \Lambda_t^\star \searrow \Longleftrightarrow std_tm_{t+1}^\star \nearrow$$

$$r_t^{\star} \searrow \Longrightarrow cov_t(m_{t+1}, m_{t+1}^{\star}) \nearrow$$

Interpretation Questions Suggestions

#### Two issues

- Backus, Foresi and Telmer (2001): Affine models and forward premium.
- ► Example 2: negative factors ⇒ negative interest rates (but low prob.);
- ► Example 4: interdependent factor model ⇒ asymmetry: one shock impacts more the foreign interest rate than the other shock, but impacts less the foreign pricing kernel.

Interpretation Questions Suggestions

# Procedure

- Quasi-MLE, using interest rate and exchange rate data.
- Paper uses two assumptions:
  - Exchange rate innovations homoskedastic:  $v_{t+1} \sim N(0, \sigma_v^2)$ .
  - Exchange rate innovations v<sub>t+1</sub> uncorrelated to interest rate residuals.
- Justifications?
  - Heteroskedasticity in FX (Andersen, Bollerslev, 1998)?
  - Currency risk premia only explained by interest rates?

Interpretation Questions Suggestions

# What drives the predictability results?

 Reference point: Clarida, Sarno, Taylor and Valente (2003): MS-VECM beats VECM by 40%.

Standard errors?

- Comparing to VARs or VECMs:
  - ▶ Is this about adding non-linearities  $(r_t)^2, (r_t^{\star})^2, ...?$
  - Or about restrictions on estimated coefficients implied by no-arbitrage?
  - $\Rightarrow$  Compare to VAR or VECM with higher moments?

Interpretation Questions Suggestions

### More moments

Variance of changes in exchange rates:

$$\frac{Q_{t+1}}{Q_t} = \frac{M_{t,t+1}^{\star}}{M_{t,t+1}}.$$

$$\sigma_{\Delta q}^2 = \sigma_{m^\star}^2 + \sigma_m^2 - 2\rho_{m^\star,m}\sigma_{m^\star}\sigma_m.$$

Campbell-Shiller tests of the EH:

$$y_{t+1}^{n-1} - y_t^n = \alpha + \beta_n \left(\frac{y_t^n - y_t^1}{n-1}\right) + \varepsilon_{t+1}.$$

Does FX data and no-arbitrage condition across countries lead to better yield predictability?

# Conclusion

- Combining bond and FX data gives much better FX out-of-sample predictability.
- Very exciting results!