

Risk Factors in Currency Markets Cont'd

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Outline

- 1 Lustig, Roussanov, and Verdelhan (2011)
 - Common Factors in Currency Returns
 - Cross-Sectional Asset Pricing

Methodology

- Linear factor models predict that average returns on a cross-section of assets can be attributed to risk premiums associated with their exposure to a **small** number of risk factors. *CAPM, 3, 4, 5*
- These factors capture **common variation** in individual asset returns.
- A principal component analysis of the currency portfolios reveals that **two factors explain more than 80% of the variation in returns on these six portfolios.**

Principal Components

Table 3
Principal components

Panel I: All Countries

about the same loading

monotonic

lowest

highest

| Portfolio | (1) | (2) | 3 | 4 | 5 | 6 |
|-----------|-------|-------|-------|-------|-------|-------|
| 1 | 0.42 | 0.43 | 0.18 | -0.15 | 0.74 | 0.20 |
| 2 | 0.38 | 0.24 | 0.15 | -0.27 | -0.61 | 0.58 |
| 3 | 0.38 | 0.29 | 0.42 | 0.12 | -0.28 | -0.71 |
| 4 | 0.38 | 0.04 | -0.35 | 0.83 | -0.03 | 0.18 |
| 5 | 0.43 | -0.08 | -0.72 | -0.44 | -0.03 | -0.30 |
| 6 | 0.45 | -0.81 | 0.35 | -0.03 | 0.11 | 0.06 |
| % Var. | 71.95 | 11.82 | 5.55 | 4.00 | 3.51 | 3.16 |

Panel II: Developed Countries

add up to 1.

| Portfolio | 1 | 2 | 3 | 4 | 5 |
|-----------|-------|-------|-------|-------|-------|
| 1 | 0.44 | 0.66 | -0.54 | -0.25 | 0.12 |
| 2 | 0.45 | 0.25 | 0.75 | 0.01 | 0.41 |
| 3 | 0.46 | 0.02 | 0.19 | 0.04 | -0.86 |
| 4 | 0.44 | -0.27 | -0.29 | 0.78 | 0.20 |
| 5 | 0.45 | -0.66 | -0.14 | -0.57 | 0.17 |
| % Var. | 78.23 | 10.11 | 4.97 | 3.49 | 3.20 |

same
informative

set 1

 Rx^1, Rx^2, \dots, Rx^6

set 2

$$PC_1 = a_1 Rx^1 + a_2 Rx^2 + \dots + a_6 Rx^6$$

$$PC_2 = b_1 Rx^1 + b_2 Rx^2 + \dots + b_6 Rx^6$$

PC3

PC4

PC5

PC6

uncorrelated.

PC₁: could explain the maximum variation of the 6 portfolios

$$Rx^1 = Z_0^{(1)} + Z_1^{(1)} PC_1 + \varepsilon^1$$

$$Rx^2 = Z_0^{(2)} + Z_1^{(2)} PC_1 + \varepsilon^2$$

$$\text{argmin}_{PC_1} \frac{\text{Var}(Z_0^{(1)}) + \text{Var}(Z_0^{(2)}) + \dots}{\text{Var}(Rx^1) + \text{Var}(Rx^2) + \dots}$$

PC₂: explain the maximum variation that is not explained by PC₁.

Interpretation

① Explain most of variation

③ Have clear interpretation

- The first principal component explains 70% of the common variation in portfolio returns, and can be interpreted as a **level factor**, since all portfolios load equally on it.
- The second principal component, that is responsible for almost 12% of common variation, can be interpreted as a **slope factor**, since portfolio loadings increase monotonically across portfolios.

Note: Average excess returns increase monotonically across portfolios \Rightarrow the second principal component is the only plausible candidate risk factor that might explain the cross-section of portfolio excess returns (since none of the other principal components exhibit monotonic variation in loadings).

Two Candidate Risk Factors *guided by PC.*

$$RX = \frac{1}{6} \sum_{i=1}^6 RX^i$$

RX

The average currency excess return, RX: is the average portfolio return of a U.S. investor who buys all foreign currencies available in the forward market (the correlation of the first principal component with RX is .99).

Carry Trade Factor: HML_{FX} *Foreign Exchange*

The difference between the return on the last portfolio and the one on the first portfolio (the correlation of the second principal component with HML_{FX} is .94). This is the return in dollars on a zero-cost strategy that goes long in the highest interest rate currencies and short in the lowest interest rate currencies.

$$HML_{FX} = RX^6 - RX^1$$

Assessing the Performance of Factors

- A linear factor model implies that the expected excess return is equal to the factor risk price times the beta of each portfolio:

$$E[R_X^j] = \underbrace{\lambda}_{\text{factor prices}} \overset{\text{transpose}}{\circledast} \underbrace{\beta^j}_{\text{portfolio betas}}$$

$$\begin{aligned} E[R_X^j] &= \beta_{RX}^j E(R_X) + \beta_{HMLFX}^j E(HML) \\ &= \underbrace{[E(R_X) \quad E(HML)]}_{\lambda} \underbrace{\begin{bmatrix} \beta_{RX}^j E(R_X) \\ \beta_{HMLFX}^j E(HML) \end{bmatrix}}_{\beta^j} \end{aligned}$$

Approach to Estimation

$$RX_{t+1}^i = \alpha^i + \beta_{RX}^i RX_{t+1} + \beta_{HML_{FX}}^i HML_{FX,t+1} + \varepsilon_{t+1}^i$$

- A two-stage ordinary least squares (OLS) estimation:
 - 1 Step 1: Run a time-series regression of returns on the factors.
 - 2 Step 2: Run a cross-sectional regression of average returns on the betas.

Model Evaluation

Table 4
Asset pricing—U.S. investor

| Panel I: Risk Prices | | | | | | | | | | | | | | |
|----------------------|-------------------|----------------|----------------|----------------|-------|------|----------|---------------------|----------------|----------------|----------------|--------|------|----------|
| | All Countries | | | | | | | Developed Countries | | | | | | |
| | λ_{HMLFX} | λ_{RX} | b_{HMLFX} | b_{RX} | R^2 | RMSE | χ^2 | λ_{HMLFX} | λ_{RX} | b_{HMLFX} | b_{RX} | R^2 | RMSE | χ^2 |
| GMM_1 | 5.50 [2.25] | 1.34 [1.85] | 0.56 [0.23] | 0.20 [0.32] | 70.11 | 0.96 | 14.39% | 3.29 [2.59] | 1.90 [2.20] | 0.29 [0.23] | 0.20 [0.23] | 64.78 | 0.64 | 45.96% |
| GMM_2 | 5.51 [2.14] | 0.40 [1.77] | 0.57 [0.22] | 0.04 [0.31] | 41.25 | 1.34 | 16.10% | 3.91 [2.52] | 3.07 [2.05] | 0.35 [0.22] | 0.32 [0.22] | -55.65 | 1.34 | 52.22% |
| FMB | 5.50 [1.79] | 1.34 [1.35] | 0.56 [0.19] | 0.20 [0.24] | 70.11 | 0.96 | 9.19% | 3.29 [1.91] | 1.90 [1.73] | 0.29 [0.17] | 0.20 [0.18] | 64.78 | 0.64 | 43.64% |
| | (1.79) | (1.35) | (0.19) | (0.24) | | | 10.20% | (1.91) | (1.73) | (0.17) | (0.18) | | | 44.25% |
| Mean | 5.08 | 1.33 | | | | | 3.14 | 1.90 | | | | | | |

Model Evaluation cont'd

Table 4
Continued

Panel II: Factor Betas

| Portfolio | All Countries | | | | | | Developed Countries | | | | | |
|-----------|-----------------|-------------------|----------------|-------|------------------|------------|---------------------|-------------------|----------------|-------|------------------|------------|
| | α_0^j | β_{HMLFX}^j | β_{RX}^j | R^2 | $\chi^2(\alpha)$ | p -value | α_0^j | β_{HMLFX}^j | β_{RX}^j | R^2 | $\chi^2(\alpha)$ | p -value |
| 1 | -0.10 [0.50] | -0.39 [0.02] | 1.05 [0.03] | 91.64 | | | 0.36 [0.53] | -0.51 [0.03] | 0.99 [0.02] | 94.31 | | |
| 2 | -1.55 [0.73] | -0.11 [0.03] | 0.94 [0.04] | 77.74 | | | -1.17 [0.85] | -0.09 [0.04] | 1.01 [0.04] | 80.69 | | |
| 3 | -0.54 [0.74] | -0.14 [0.03] | 0.96 [0.04] | 76.72 | | | 0.62 [0.79] | -0.00 [0.03] | 1.04 [0.03] | 86.50 | | |
| 4 | 1.51 [0.77] | -0.01 [0.03] | 0.95 [0.05] | 75.36 | | | -0.17 [0.85] | 0.12 [0.03] | 0.97 [0.04] | 82.84 | | |
| 5 | 0.78 [0.82] | 0.04 [0.03] | 1.06 [0.05] | 76.41 | | | 0.36 [0.53] | 0.49 [0.03] | 0.99 [0.02] | 94.32 | | |
| 6 | -0.10 [0.50] | 0.61 [0.02] | 1.05 [0.03] | 93.84 | | | | | | | | |
| All | | | | | 6.79 | 34.05% | | | | | 63 | 75.64% |

annualized

standard error

significant

monotonically
increaseclose
to 1.

large

 \Rightarrow Model performs wellHigh interest rate
currency is highly
exposed to value risk.

small

monotonic
increasingclose
to 1

high