

# International Macro-Finance & Long-Run Risks: an overview

**Ric Colacito**



THE UNIVERSITY  
*of* NORTH CAROLINA  
*at* CHAPEL HILL

# The FX volatility puzzle




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- Solution: a model with a large international correlation of long-term growth prospects
- Reading list:
  - ↪ Bansal and Yaron, JF 2004 
  - ↪ Brandt, Cochrane, and Santa Clara, JME 2006 
  - ↪ Colacito and Croce, JPE 2011 

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## Setup of the economy

- Endowment economy
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- Two country specific goods
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- Epstein, Zin and Weil preferences:

$$U_t^i = \left\{ (1 - \delta)(C_t^i)^{\frac{1-\gamma}{\theta}} + \delta [E_t(U_{t+1}^i)^{1-\gamma}]^{\frac{1}{\theta}} \right\}^{\frac{\theta}{1-\gamma}}, \forall i \in \{h, f\}$$

where

$$\theta = \frac{1 - \gamma}{1 - 1/\psi}$$

## Dynamics of consumption

Based on Bansal and Yaron (JF, 2004):

$$\Delta c_t^i = \mu_c + x_{t-1}^i + \sigma \underbrace{\varepsilon_{c,t}^i}_{\text{short-run shocks}}$$

$$x_t^i = \rho x_{t-1}^i + \sigma \varphi_e \underbrace{\varepsilon_{x,t}^i}_{\text{long-run shocks}}, \quad \forall i \in \{h, f\}$$

Shocks correlations:

- orthogonal within each country
- correlated across countries
  - ↪  $\rho_c = \text{corr}(\varepsilon_{c,t}^h, \varepsilon_{c,t}^f)$
  - ↪  $\rho_x = \text{corr}(\varepsilon_{x,t}^h, \varepsilon_{x,t}^f)$

## Stochastic discount factors ( $\psi = 1$ )

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$$m_{t+1}^i = \log \delta - \Delta c_{t+1}^i + \log \frac{\exp\{(1 - \gamma)U_{t+1}^i\}}{E_t [\exp\{(1 - \gamma)U_{t+1}^i\}]}$$

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- Combine:
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- Economic intuition:

↪ short-run: slow diffusion of technologies and ideas across countries

↪ long-run: countries converge to the same technological frontier

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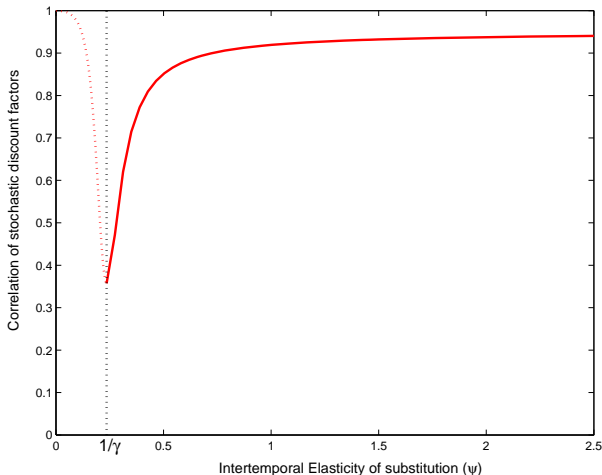
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- Agents are extremely sensitive to long-run news
- Long-run risks should be very correlated to replicate FX volatility

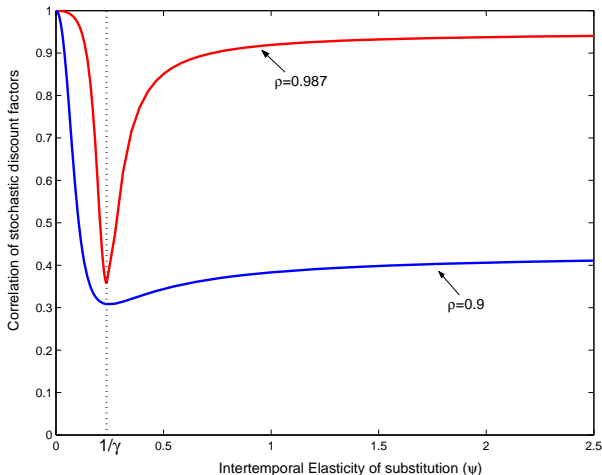
# SDF Correlation

$$m_{t+1}^i = \theta \log \delta - \frac{1}{\psi} x_t^i - \gamma \sigma \varepsilon_{c,t+1}^i + \frac{\delta(1-\gamma\psi)}{\psi(1-\rho\delta)} \sigma \varphi_e \varepsilon_{x,t+1}^i$$



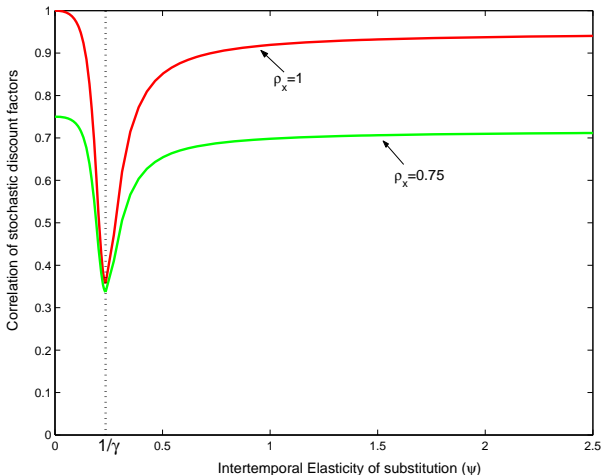
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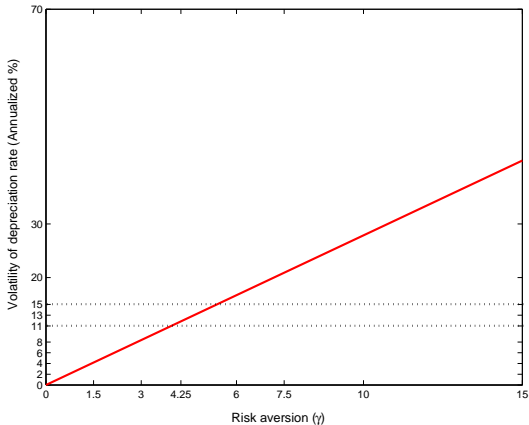


## FX Volatility

$$\text{Var}\left(\frac{e_{t+1}}{e_t}\right) = \frac{2(1-\rho_x)}{\psi^2} \left\{ \frac{1}{1-\rho^2} + \left[ \frac{\delta(1-\gamma\psi)}{(1-\rho\delta)} \right]^2 \right\} \varphi_e^2 \sigma^2 + 2\gamma^2(1-\rho_c)\sigma^2$$

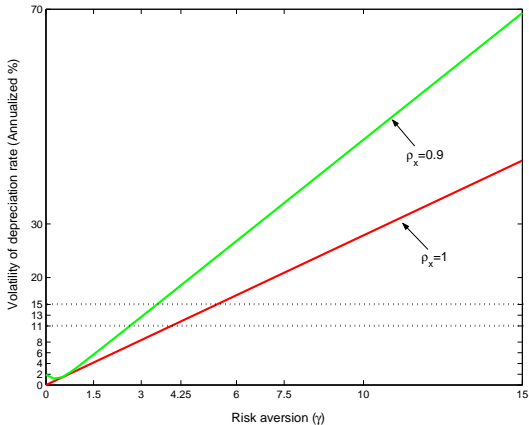
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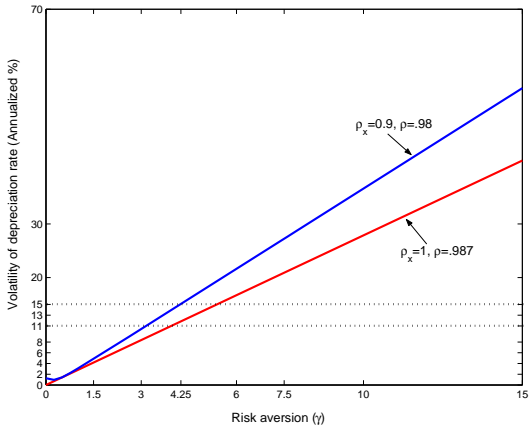
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## Estimation

Use predictive regressions

$$\Delta c_t^i = \mu_c + \underbrace{Z_{t-1}^i \cdot \beta^i}_{x_{t-1}^i} + \varepsilon_{c,t}^i$$

where  $Z_t^i = [pd_t^i, r_{f,t}^i, \Delta c_t^i, \Delta cy_t^i, default_t^i]$ ,  $\forall i \in \{h, f\}$ .

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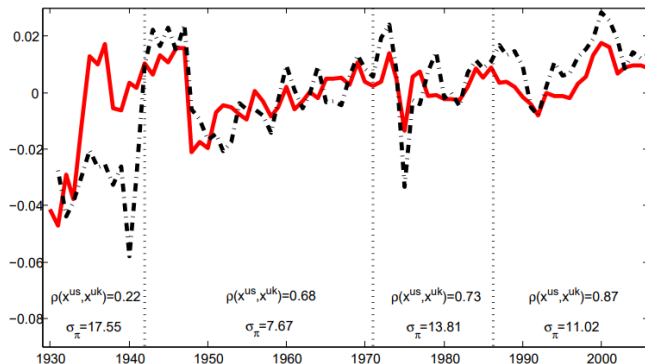
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	<i>F - stat</i>		<i>R</i> <sup>2</sup>		$\rho_x$		<i>corr</i> ( $x^{US}, x^{UK}$ )
	US	UK	US	UK	US	UK	
Pd and risk-free	6.641 (0.002)	3.476 (0.036)	0.145	0.086	0.768 (0.076)	0.787 (0.074)	0.579 [0.137, 0.896]
All predictive variables	6.585 (0.000)	6.852 (0.000)	0.315	0.278	0.672 (0.074)	0.759 (0.074)	0.758 [0.531, 0.922]
Pd only	7.299 (0.008)	6.013 (0.016)	0.086	0.076	0.885 (0.065)	0.726 (0.099)	0.849 [0.762, 0.941]
VAR	11.719 (0.000)	4.204 (0.008)	0.325	0.149	0.869 (0.058)	0.765 (0.075)	0.717 [0.172, 0.953]

## FX vol and correlation of predictive components



FX tend to be smoother when predictive components are more correlated

# Estimating preference parameters

	Conditional Estimation			Joint Estimation		
	$P/D$	$P/D, R_f$	All	$P/D$	$P/D, R_f$	All
$\psi$	<b>4.094</b> [ 0.398, 6.739 ]	<b>1.371</b> [ 0.391, 1.792 ]	<b>1.276</b> [ 0.404, 1.992 ]	<b>2.781</b> [ 0.243, 2.971 ]	<b>2.719</b> [ 0.288, 3.030 ]	<b>2.001</b> [ 0.274, 3.099 ]
$\gamma$	<b>4.359</b> [ 3.181, 9.575 ]	<b>3.402</b> [ 2.282, 7.376 ]	<b>2.936</b> [ 2.144, 6.339 ]	<b>3.387</b> [ 1.014, 10.267 ]	<b>3.225</b> [ 1.090, 9.509 ]	<b>4.147</b> [ 1.167, 10.802 ]
$\rho_x$	-	-	-	<b>0.997</b> [ 0.427, 0.999 ]	<b>0.997</b> [ 0.413, 0.999 ]	<b>0.988</b> [ 0.662, 0.999 ]
$\rho(x^{US}, x^{UK})$	-	-	-	<b>0.879</b> [ 0.618, 0.936 ]	<b>0.867</b> [ 0.582, 0.939 ]	<b>0.830</b> [ 0.471, 0.963 ]
Wald-stat p-value	-	-	-	0.000	0.000	0.000
$\gamma - 1/\psi$ p-value	0.000	0.002	0.008	0.000	0.001	0.000
J-stat p-value	0.000	0.000	0.000	0.035	0.000	0.011

- GMM estimation with Euler equation restrictions and first two moments of FX
- Low risk aversion
- IES around 2