

Les 3èmes
« TOULOUSE LECTURES IN ECONOMICS »

9-10-11 MAI 2005

Challenges in the Measurement of Long Run Risk

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Recursive utility and risk

Under a Cobb-Douglas specification ($\rho = 1$), CES recursion is

$$V_t = (C_t)^{(1-\beta)} \mathcal{R}_t(V_{t+1})^\beta.$$

Stochastic discount factor:

$$s_{t+1,t} \equiv \log S_{t+1,t} = -\delta - \gamma(L)w_{t+1} - \mu_c + (1-\alpha)\gamma(\beta)w_{t+1} - \frac{(1-\alpha)^2\gamma(\beta) \cdot \gamma(\beta)}{2}$$

where $\beta = \exp(-\delta)$ and

$$\gamma(z) = \sum_{j=0}^{\infty} \gamma_j z^j$$

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where $\beta = \exp(-\delta)$ and

$$\gamma(z) = \sum_{j=0}^{\infty} \gamma_j z^j$$

- $\gamma(0) = \gamma_0$ - Lucas (1978)-Breedon (1979) adjustment for risk
- $(\alpha - 1)\gamma(\beta)$ - Kreps and Porteus (1978) recursive utility adjustment for risk - featured by Bansal and Yaron (2004)
- remainder contributes to the risk free rate

Measured Consumption Dynamics

$$A_0 y_t + A_1 y_{t-1} + A_2 y_{t-2} + \dots + A_\ell y_{t-\ell} + B_0 = w_t,$$

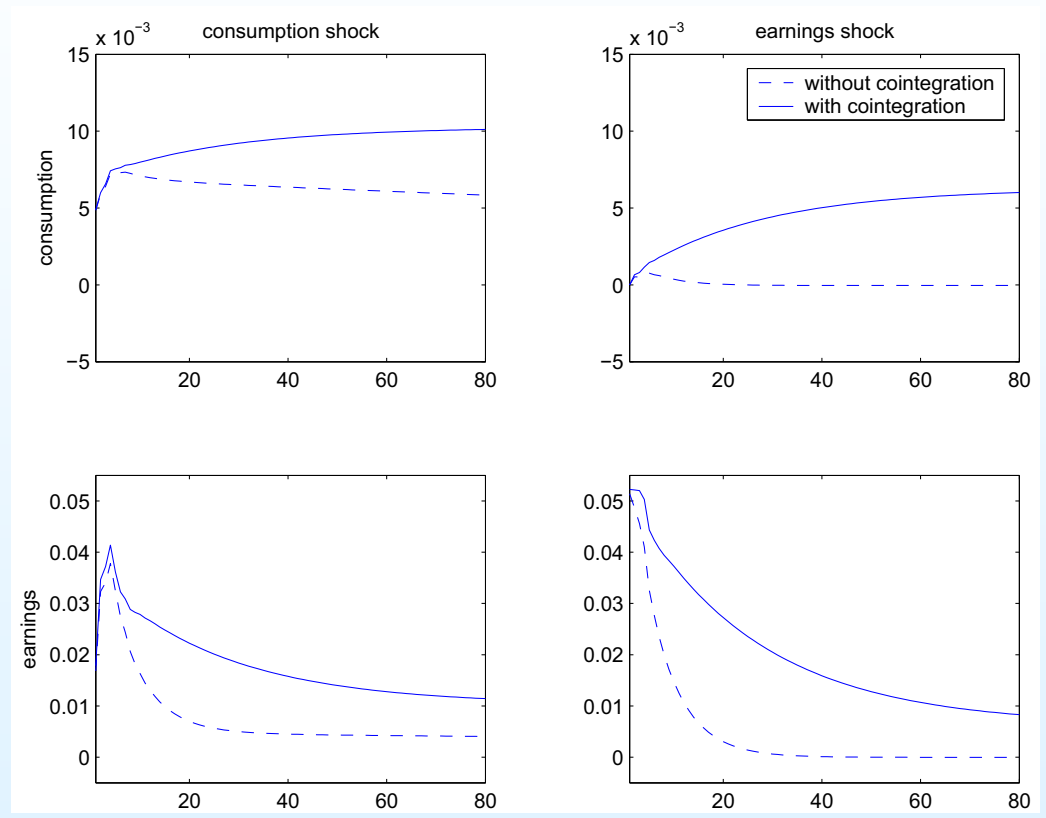
- Shock vector w_t has mean zero and covariance matrix I and normally distributed. A_0 lower triangular with positive entries on the diagonals.

$$A(z) \doteq A_0 + A_1 z + A_2 z^2 + \dots + A_\ell z^\ell.$$

where $A(z)$ is nonsingular for $|z| < 1$.

- Log consumption is the first coordinate of y_t and corporate earnings is the second (used as an investor information variable);
- Cointegration: $A(1) = a_1 \begin{bmatrix} 1 & -\lambda \end{bmatrix}$. $\lambda = 1$ balanced growth restriction.

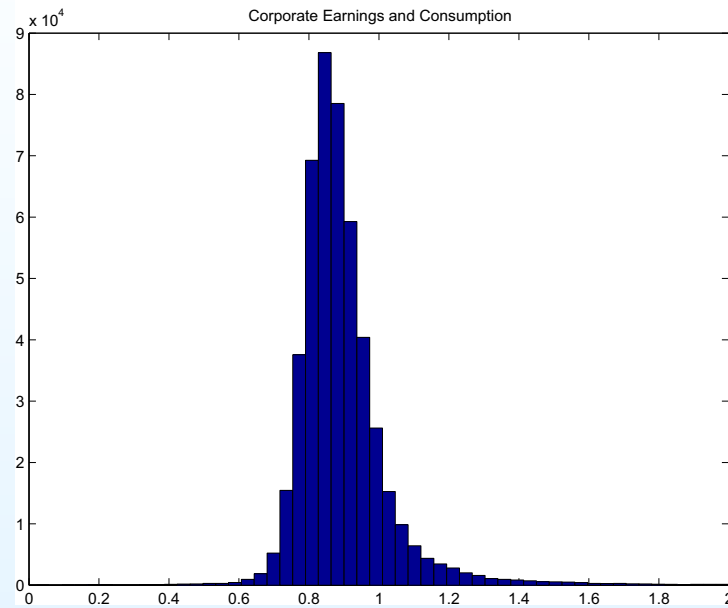
Consumption-Earnings VAR



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Approximate Probabilities for λ



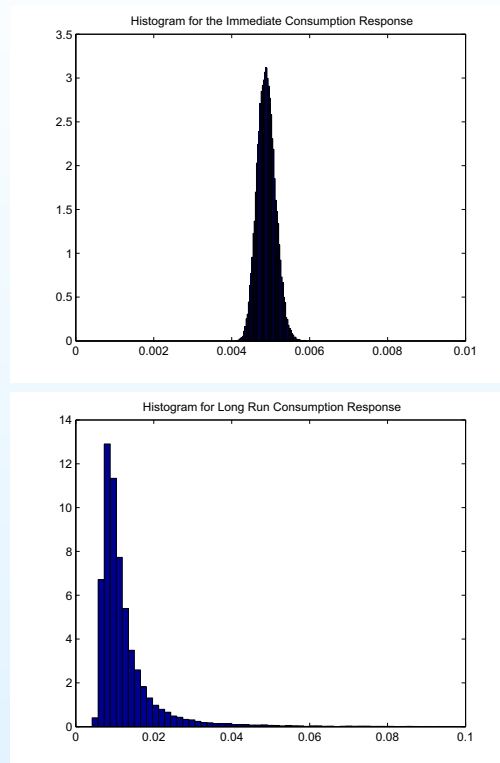
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Statistical Accuracy of Responses



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Three operators

1. One-period valuation-growth operator:

$$\mathcal{P}\psi(x) = E [\exp (s_{t+1,t} + \zeta + M_{t+1} - M_t) \psi(x_{t+1}) | x_t = x].$$

$\{M_t\}$ is the martingale component for cash flow.

2. One-period growth operator (abstracts from valuation)

$$\mathcal{G}\psi(x) = E [\exp (\zeta + M_{t+1} - M_t) \psi(x_{t+1}) | x_t = x].$$

3. One-period valuation operator (abstracts from growth)

$$\mathcal{P}^f \psi(x) = E [\exp (s_{t+1,t}) \psi(x_{t+1}) | x_t = x].$$

Long run value accounting

Three eigenvalue problems:

1. Solve $\mathcal{P}\phi^* = \exp(-\nu)\phi^*$.
 ν asymptotic rate of decay in value.
2. $\mathcal{G}\phi^+ = \exp(\epsilon)\phi^+$.
 ϵ asymptotic growth rate of cash flow.
 $\nu + \epsilon$ is an expected rate of return.
3. Solve $\mathcal{P}^f\phi_f = \exp(-\nu_f)\phi_f$.
 $\nu + \epsilon - \nu_f$ expected excess rate of return.

Change martingales trace out long run risk-return relation.

Example Economy

Cash flow growth:

$$d_t - d_{t-1} = \mu_d + \iota(L)w_t$$

where $\iota(z) = \sum_{j=0}^{\infty} \iota_j z^j$.

- Construct martingale M_t to have increment $\iota(1)w_t$.
- Deterministic growth rate is μ_d .
- Growth eigenvalue is $\epsilon = \mu_d + \frac{1}{2}|\iota(1)|^2$

Stochastic Discount Factor

$$s_{t+1,t} = -\delta - \gamma(L)w_{t+1} - \mu_c + (1 - \alpha)\gamma(\beta)w_{t+1} - \frac{(1 - \alpha)^2\gamma(\beta) \cdot \gamma(\beta)}{2}$$

Dominant eigenvalue for valuation:

$$\nu = \delta - \mu_d - \frac{1}{2} | -\gamma(1) + (1 - \alpha)\gamma(\beta) + \iota(1) |^2$$

Rate of return:

$$\nu + \epsilon = \delta - \frac{1}{2} | -\gamma(1) + (1 - \alpha)\gamma(\beta) |^2 + [\gamma(1) + (\alpha - 1)\gamma(\beta)] \cdot \iota(1)$$

Excess return:

$$[\gamma(1) + (\alpha - 1)\gamma(\beta)] \cdot \iota(1)$$

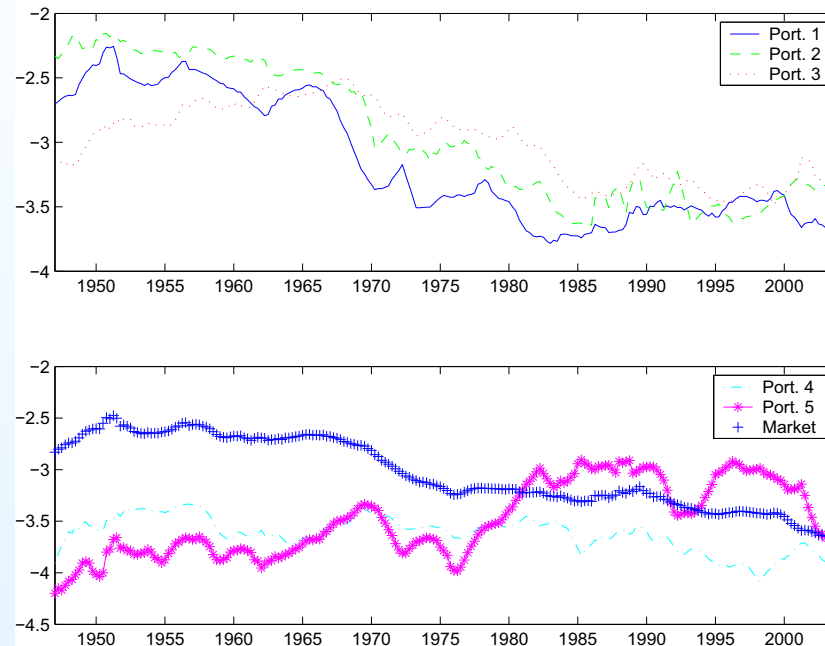
Robustness and Intertemporal Substitution

- Robustness without hidden states $\theta = \frac{\beta}{\alpha-1}$. Tallarini (1998)
- Intertemporal substitution - compute derivatives of dominant eigenvalue with respect to ρ .

Return Properties of Portfolios

| | Portfolio | | | | | Market |
|------------------------------|-----------|------|------|------|-------|--------|
| | 1 | 2 | 3 | 4 | 5 | |
| Avg. Return (%) | 6.48 | 6.88 | 8.90 | 9.32 | 11.02 | 7.23 |
| Std. Return (%) | 18.8 | 16.4 | 14.8 | 15.8 | 17.8 | 16.5 |
| Avg. B/M | 0.32 | 0.62 | 0.84 | 1.12 | 2.00 | 0.79 |
| Sharpe Ratio | 0.18 | 0.20 | 0.28 | 0.28 | 0.30 | 0.21 |
| Correlation with Consumption | 0.20 | 0.18 | 0.20 | 0.20 | 0.21 | 0.20 |

Implied Cash Flows



Tail Returns

| $\alpha = 1$ | | | | |
|--------------|--------|---------------|-------------------|--------------------------|
| Portfolio | Return | Excess Return | Return Derivative | Excess Return Derivative |
| 1 | 6.48 | -.06 | 3.51 | .00 |
| 5 | 6.78 | .22 | 3.53 | .02 |
| market | 6.60 | .06 | 3.52 | .00 |
| long bond | 6.54 | .00 | 3.51 | .00 |
| $\alpha = 5$ | | | | |
| Portfolio | Return | Excess Return | Return Derivative | Excess Return Derivative |
| 1 | 6.11 | -.27 | 3.48 | .03 |
| 5 | 7.43 | 1.03 | 3.35 | -.10 |
| market | 6.69 | .30 | 3.42 | -.03 |
| long bond | 6.39 | .00 | 3.45 | .00 |

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Tail Returns

 $\alpha = 10$

| Portfolio | Return | Excess Return | Return Derivative | Excess Return Derivative |
|-----------|--------|---------------|-------------------|--------------------------|
| 1 | 5.65 | -0.54 | 3.44 | .07 |
| 5 | 8.25 | 2.05 | 3.12 | -.25 |
| market | 6.79 | 0.60 | 3.29 | -.08 |
| long bond | 6.19 | 0.00 | 3.37 | .00 |

 $\alpha = 20$

| Portfolio | Return | Excess Return | Return Derivative | Excess Return Derivative |
|-----------|--------|---------------|-------------------|--------------------------|
| 1 | 4.74 | -1.72 | 3.34 | .14 |
| 5 | 9.90 | 4.09 | 2.70 | -.50 |
| market | 7.00 | 1.19 | 3.06 | -.15 |
| long bond | 5.81 | 0.00 | 3.21 | .00 |

VAR Growth Models

We append a dividend equation

$$A_0^* y_t^* + A_1^* y_{t-1} + A_2^* y_{t-2} + \dots + A_\ell^* y_{t-\ell} + B_0^* + B_1^* t = w_t^* ,$$

to VAR. The vector of inputs is

$$y_t^* \doteq \begin{bmatrix} y_t \\ d_t \end{bmatrix} = \begin{bmatrix} c_t \\ e_t \\ d_t \end{bmatrix}$$

and the shock w_t^* is scalar with mean zero and unit variance.

Three Alternative Models

1. Cointegration: $B_1^* = 0$ suppose the vector $A^*(1)$ is orthogonal to $\begin{bmatrix} 1 & 1 & \lambda^* \end{bmatrix}$ where

$$A^*(z) \doteq \sum_{j=0}^{\ell} A_j^* z^j.$$

For this model, the vector π is

$$\pi = \iota(1) = \lambda^* \gamma(1)$$

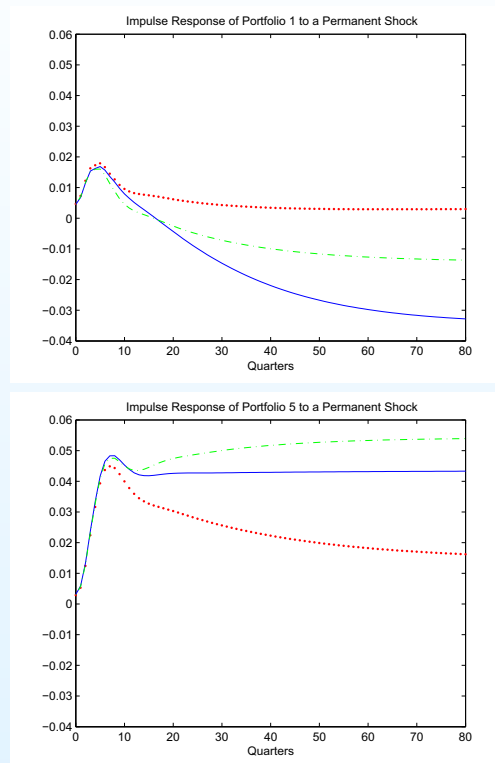
and $\zeta = \mu_d = \lambda^* \mu_c$.

2. Time trend: $B_1^* \neq 0$ but same rank restriction. Leaves μ_d unrestricted.
3. Unit root: $B_1^* = 0$ and $A^*(1) = \begin{bmatrix} a_1^* & -a_1^* & 0 \end{bmatrix}$. New dividend shock is permanent.

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Estimated cash flow responses to macro shocks



Long Run Responses

| No Time Trends | | | | | |
|----------------|----------|------|------|------|------|
| Portfolio | Quantile | | | | |
| | .05 | .25 | .5 | .75 | .95 |
| 1 | -1.35 | 0.08 | 0.35 | 0.53 | 0.83 |
| 2 | 0.18 | 0.47 | 0.66 | 0.98 | 2.46 |
| 3 | 0.54 | 0.69 | 0.78 | 0.88 | 1.09 |
| 4 | 0.55 | 0.65 | 0.70 | 0.77 | 0.89 |
| 5 | 0.65 | 1.21 | 1.44 | 1.71 | 3.00 |
| market | 0.36 | 0.44 | 0.49 | 0.53 | 0.61 |

Long Run Responses

| Time Trend Included | | | | | |
|---------------------|----------|-------|-------|-------|-------|
| Portfolio | Quantile | | | | |
| | .05 | .25 | .5 | .75 | .95 |
| 1 | -6.24 | -4.26 | -3.42 | -2.72 | -1.77 |
| 2 | -16.47 | -7.28 | -4.70 | -3.15 | -1.52 |
| 3 | -0.64 | 1.21 | 2.09 | 2.84 | 3.87 |
| 4 | 1.10 | 1.70 | 2.07 | 2.44 | 3.05 |
| 5 | -1.66 | 1.66 | 3.87 | 7.32 | 21.18 |
| market | -1.60 | -0.43 | 0.10 | 0.53 | 1.12 |

Long Run Responses

| Portfolio | Dividend Growth Rate | | | | |
|-----------|----------------------|-------|-------|-------|------|
| | Quantile | | | | |
| | .05 | .25 | .5 | .75 | .95 |
| 1 | -4.84 | -2.61 | -1.35 | -0.20 | 1.60 |
| 2 | -2.13 | -0.68 | 0.18 | 1.03 | 2.43 |
| 3 | 0.34 | 1.82 | 2.97 | 4.32 | 7.05 |
| 4 | 1.82 | 3.05 | 3.97 | 5.01 | 6.94 |
| 5 | 1.42 | 3.64 | 5.18 | 6.87 | 9.93 |
| market | -0.20 | 0.84 | 1.49 | 2.20 | 3.49 |

Heteroskedasticity

Consider an environment with stationary linear dynamics:

$$x_{t+1} = Gx_t + Hw_{t+1}$$

where w_{t+1} is a multivariate normal with mean zero and covariance matrix I . Introduce a stochastic discount factor of the form $\exp(s_{t+1})$ where

$$s_{t+1,t} = \frac{1}{2}w_{t+1}'\Theta_0w_{t+1} + x_t'\Theta_1w_{t+1} + \frac{1}{2}x_t'\Theta_2x_t + \Lambda_0w_{t+1} + \Lambda_1x_t + \lambda.$$

Find a dominant eigenfunction of the form:

$$\phi(x) = \exp\left(-\frac{1}{2}x'P^*x - Q^*x\right)$$

Lettau-Wachter consider heteroskedastic cash flows and constant interest rates. Eigenfunction is the exponential of a linear function of the state. Selects volatility contribution to risk premia as a source of long run risk

Why Heteroskedasticity?

Let

$$c_{t+1} - c_t = \mu_c + U_c x_t + x_t' V_c w_{t+1} + \gamma_0 w_{t+1}.$$

Consider again the recursive utility model:

$$v_t = \frac{\beta}{1 - \alpha} \log E (\exp [(1 - \alpha)(v_{t+1} + c_{t+1} - c_t)] | \mathcal{F}_t)$$

- Value function

$$v^* = -\frac{1}{2} x^{*'} \Omega^* x^* + \Delta^* x^* + \omega^*$$

- Worst case model:

$$m_{t+1} = \frac{\exp \left(-\frac{v_{t+1} + c_{t+1} - c_t}{\theta} \right)}{E \left[\exp \left(-\frac{v_{t+1} + c_{t+1} - c_t}{\theta} \right) | x_t \right]}$$

Alternative Formulations of Consumption Volatility

- Tauchen adopts an alternative continuous state specification for consumption volatility.
- Lettau, Ludvigson and Wachter adopt a regime shift formulation.
- Robustness gives a different way to assign probabilities for asset pricing.

Hidden State Formulation

Let log consumption evolve as:

$$c_{t+1} - c_t = Gx_t + Hw_{t+1}$$

where

$$x_{t+1} = Ax_t + Cw_{t+1}$$

and

$$x_t = \begin{bmatrix} y_t \\ z_t \end{bmatrix}.$$

Examples:

1. hidden predictor variables for log consumption
2. long run growth rate is estimated (constant hidden state)

Basic Idea

Two robustness recursions:

1. Allow for misspecified dynamics as before using m_{t+1} conditioned on x_t .
2. Allow for robustness in the probabilities used for averaging over the hidden states as in prior sensitivity analysis in Bayesian statistics and econometrics - perturb normal distributions for hidden states

Observations

1. hidden states can be time invariant - consumption growth rate
2. exploit tools for solving hidden state Markov chain models - Kalman filter
3. minimizing agent has an informational advantage - observe z_t
4. both "games" easy to solve

Multiple Models

1. Model 1

$$y_{t+1} = A_1 y_t + C_1 w_{t+1}$$

2. Model 2

$$y_{t+1} = A_2 y_t + C_2 w_{t+1}$$

Observations

- Hidden state is a model indicator.
- Make robust adjustments to posterior probabilities used to mix models in addition to a robustness adjustment for each model

Technology Processes with Hidden Growth States

- Technology shock evolves as:

$$y_{t+1} - y_t = \kappa z_t + \sigma_y w_{t+1}$$

where $\{z_t\}$ is a hidden growth regime process that evolves as a Markov chain.

- Given adjustment costs in investment and lack of perfect knowledge of the technology growth state, consumption response will be sluggish
- Make robust adjustments to the shock process conditioned on z_t and robust adjustments to the probabilities assigned to hidden states.
- Additional sources of "risk premia".