

TERM STRUCTURE OF RISK PRICES AND EXPOSURES IN NONLINEAR MODELS

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The manner in which risk operates upon time preference will differ, among other things, according to the *particular periods in the future to which the risk applies*.

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There are several alternative ways in which one may approach the *impulse problem* One way which I believe is particularly fruitful and promising is to study what would become of the solution of a determinate dynamic system if it were *exposed to a stream of erratic shocks* that constantly upsets the continuous evolution, and by so doing introduces into the system the energy necessary to maintain the swings.

Ragnar Frisch (1933)

Idea

- quantify **cash-flow exposures** to shocks over alternative investment horizons;
- and the associated **compensations to investors** holding these cash flows

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Shock elasticities: refinement of the analysis of asset pricing models

- **shock-exposure elasticity** – sensitivity of **expected payoffs** to a shock
 - risk quantification
- **shock-price elasticity** – sensitivity of **expected returns** to shock exposure
 - risk pricing

Dynamic value decomposition

- expected returns and payoffs \implies shock elasticities

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Alternative payoff horizons

- term structure of risk

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Shock elasticities

- an alternative to log-linearization methods in asset pricing
- asset pricing counterparts to **nonlinear impulse response functions**.
- framework explicitly takes into account:
 - nonlinearity (stochastic volatility, regime shifts, ...)
 - growth and discounting (compounding over time, permanent components)

Empirical evidence

- growing literature on term structure of risky returns
- new evidence to help us understand differences between existing models and how to distinguish them

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Theoretical modeling

- models make assumptions about the temporal composition of risk
 - transitory vs permanent components
 - substantial role of the permanent component for pricing
- methods for formal description of these components

Shock elasticities

- One-period pricing;
- Elasticities over multiple periods;
- Martingale components in stochastic discount factors

Applications

- Exponential-quadratic framework solvable in closed form;
 - Second-order perturbations of DSGE models (Dynare implementation);
 - Model of intangible capital [Ai, Croce and Li (2012)]
 - Model with financial constraints [Gertler and Kiyotaki (2011)]
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- Suppose X is first-order Markov, and $W \sim N(0, I)$ iid

$$X_{t+1} = \psi(X_t, W_{t+1})$$

- A **multiplicative functional** M is a conditionally Gaussian model

$$\log M_{t+1} - \log M_t = \kappa(X_t, W_{t+1})$$

- Captures growth and decay.

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- Captures growth and decay.
- Examples of multiplicative functionals
 - stochastic discount factor S
 - macroeconomic growth functional G (consumption, priced cash flows, ...)

- Given state X_0 , suppose that

$$\log G_1 = \beta_g(X_0) + \alpha_g(X_0) \cdot W_1$$

$$\log S_1 = \beta_s(X_0) + \alpha_s(X_0) \cdot W_1$$

- One-period return

$$R_1 = \frac{G_1}{E[S_1 G_1 | X_0]}$$

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- Logarithm of the expected return

$$\begin{aligned} \log E[R_1 | X_0 = x] &= \log E[G_1 | X_0 = x] - \log E[S_1 G_1 | X_0 = x] = \\ &= -\beta_s(x) - \frac{1}{2} |\alpha_s(x)|^2 - \alpha_g(x) \cdot \alpha_s(x) \end{aligned}$$

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- $-\alpha_s(x)$ is the risk price vector for exposure to components of W_1 .
- Asset pricing puzzle:** Modeled versions of $|\alpha_s|$ are too small for 'plausible' values of risk aversion.

- Consider a parameterized family of **perturbations**

$$\log H_1(r) = -\frac{1}{2}r^2 |\eta(X_0)|^2 + r\eta(X_0) \cdot W_1$$

- exposure direction η , scaling r , and $E[H_1(r) | X_0] = 1$

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- exposure direction η , scaling r , and $E[H_1(r) | X_0] = 1$
- Form $G_1 H_1(r)$:

$$\log G_1 + \log H_1(r) = \beta_g(X_0) - \frac{1}{2}r^2 |\eta(X_0)|^2 + \underbrace{[\alpha_g(X_0) + r\eta(X_0)]}_{\text{new shock exposure}} \cdot W_1$$

- Parameterized family of payoffs to be priced.

Compute expected returns

$$\log E[G_1 H_1(r) | X_0 = x] - \log E[S_1 G_1 H_1(r) | X_0 = x]$$

Compute expected returns and differentiate

$$\frac{d}{dr} \log E[G_1 H_1(r) | X_0 = x] - \log E[S_1 G_1 H_1(r) | X_0 = x] \Big|_{r=0}$$

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Shock elasticities

1. shock-exposure elasticity

$$\varepsilon_g(x) = \left. \frac{d}{dr} \log E[G_1 H_1(r) | X_0 = x] \right|_{r=0} = \alpha_g(x) \cdot \eta(x)$$

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2. shock-cost elasticity

$$\varepsilon_v(x) = \frac{d}{dr} \log E[S_1 G_1 H_1(r) | X_0 = x] \Big|_{r=0} = [\alpha_s(x) + \alpha_g(x)] \cdot \eta(x)$$

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3. shock-price elasticity

$$\varepsilon_p(x) = \varepsilon_g(x) - \varepsilon_v(x) = -\alpha_s(x) \cdot \eta(x)$$

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- Time- t payoff G_t (zero-coupon equity), stochastic discount factor S_t
- Perturb payoff in **period 1**: $G_t H_1(r)$
- Expected return

$$\frac{1}{t} \log E[G_t H_1(r) | X_0 = x] - \frac{1}{t} \log E[S_t G_t H_1(r) | X_0 = x]$$

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- Form shock elasticities:
 1. Shock-**exposure** elasticity

$$\varepsilon_g(x, t) = \left. \frac{d}{dr} \log E[G_t H_1(r) | X_0 = x] \right|_{r=0}$$

2. Shock-**value** elasticity

$$\varepsilon_v(x, t) = \left. \frac{d}{dr} \log E[S_t G_t H_1(r) | X_0 = x] \right|_{r=0}$$

3. Shock-**price** elasticity

$$\varepsilon_p(x, t) = \varepsilon_g(x, t) - \varepsilon_v(x, t)$$

Let M be a multiplicative functional (either G or SG):

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Observations

- Log-linear framework:
 - $\varepsilon(x, t)$ recovers the impulse response function for $\log M$ in response to a shock $\eta(x) \cdot W_1$
 - **shock-exposure elasticity** $\varepsilon_g(x, t)$ reflects the IRF for $\log G$
 - **shock-price elasticity** $\varepsilon_p(x, t)$ reflects the IRF for $-\log S$

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- With stochastic volatility and other sources of nonlinearity, $\varepsilon_p(x, t)$ will also depend on the choice of G

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What do long-term pricing implications depend on?

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Multiplicative structure of M allows the extraction of a martingale component

- Hansen and Scheinkman (2009), Alvarez and Jermann (2005)
- long-horizon elasticity limits driven by permanent components of M

- We want to pin down long-horizon limits (role of permanent shocks).
- Consider the eigenvalue problem (M_0 normalized to one)

$$E[M_1 e(X_1) \mid X_0 = x] = \exp(\eta) e(x)$$

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We obtain a martingale

$$\tilde{M}_t = M_t \exp(-\eta t) \frac{e(X_t)}{e(X_0)}$$

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- \tilde{M} is a martingale
- $\exp(\eta t)$ is a deterministic drift
- $e(X_t)$ is a stationary (eigen)function of the Markov state X_t

- Martingale factorization

$$M_t = \exp(\eta t) \tilde{M}_t \frac{e(X_0)}{e(X_t)}$$

- Thus (using $\hat{e}(x) = 1/e(x)$)

$$\varepsilon(x, t) = \eta(x) \cdot \frac{E[M_t W_1 | X_0 = x]}{E[M_t | X_0 = x]} = \eta(x) \cdot \frac{\tilde{E}[\hat{e}(X_t) W_1 | X_0 = x]}{\tilde{E}[\hat{e}(X_t) | X_0 = x]}$$

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- Long-horizon limit

$$\varepsilon(x, \infty) = \eta(x) \cdot \tilde{E}[W_1 | X_0 = x]$$

Short-horizon limit $\varepsilon(x, 1)$

- corresponds to the **local exposure and price of risk** in the continuous-time limit

Long-horizon limit $\varepsilon(x, \infty)$

- pricing determined by the martingale components in cash flows and SDFs

Shock elasticities fill the rest

- **term structure of risk**

Martingale component in the SDF generated by

- permanent shocks to investors' consumption
- nonseparabilities in preferences (Epstein-Zin)
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Result

$$L_t \left(\frac{\tilde{S}_{t+1}}{\tilde{S}_t} \right) \geq E_t \log R_{t,t+1} - E_t \log R_{t,t+1}^\infty$$

- $R_{t,t+1}$ – any one-period return
- $R_{t,t+1}^\infty$ – one-period return on the long-horizon bond
- $L_t(Z_{t+1}) = \log E_t [Z_{t+1}] - E_t [\log Z_{t+1}]$ – entropy of random variable Z_{t+1}

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Triangular state vector system:

$$\begin{aligned}
 X_{1,t+1} &= \Delta_{10} + \Delta_{11}X_{1,t} + \Lambda_{10}W_{t+1} \\
 X_{2,t+1} &= \Delta_{20} + \Delta_{21}X_{1,t} + \Delta_{22}X_{2,t} + \Delta_{23}(X_{1,t} \otimes X_{1,t}) + \\
 &\quad + \Lambda_{20}W_{t+1} + \Lambda_{21}(X_{1,t} \otimes W_{t+1}) + \Lambda_{23}(W_{t+1} \otimes W_{t+1})
 \end{aligned}$$

- Stable dynamics if Δ_{11} and Δ_{22} have stable eigenvalues.
- Structure allows for stochastic volatility through $X_{1,t} \otimes W_{t+1}$

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A class of quadratic functions for increments of multiplicative functionals

$$\begin{aligned} \log M_{t+1} - \log M_t &= \Gamma_0 + \Gamma_1 X_t + \Gamma_2 (X_{1,t} \otimes X_{1,t}) + \\ &\quad + \Theta_0 W_{t+1} + \Theta_1 (X_{1,t} \otimes W_{t+1}) + \Theta_2 (W_{t+1} \otimes W_{t+1}) \end{aligned}$$

- Use to model stochastic discount factors and growth functionals.

The framework allows for quasi-analytical formulas for conditional expectations of multiplicative functionals and for elasticities.

- Start with

$$\log f(x) = \phi + \Phi x + \frac{1}{2}(x_1)' \Psi(x_1)$$

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$$\log f(x) = \phi + \Phi x + \frac{1}{2}(x_1)' \Psi (x_1)$$

- Then

$$\begin{aligned} \log E \left[\left(\frac{M_{t+1}}{M_t} \right) f(X_{t+1}) | X_t = x \right] &= \phi^* + \Phi^* x + \frac{1}{2}(x_1)' \Psi^* (x_1) = \\ &= \log f^*(x) \end{aligned}$$

- Multiperiod conditional expectations calculated by iterating on the coefficient mapping.
- Series expansion (perturbation) framework as a special case.

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Value premium

- High book-to-market (B/M) firms (**value stocks**) have high expected returns relative to low B/M firms (**growth stocks**) [Fama and French (1992, 1995)]
- Does this reflect intangible risk exposure? [Hansen, Heaton and Li (2005)].

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Use a model by Ai, Croce, Li (2012)

- distinction between **physical** and **intangible capital**
- growth firms have more intangible capital that is a source of new investment projects
- new vintages of capital are less exposed to aggregate risk than old vintages
- growth firms are less risky, and thus earn lower expected returns

Aggregate output in the economy

$$Y_t = (K_t)^\lambda (A_t N_t)^{1-\lambda} = C_t + I_t + I_t^* \quad N_t = 1$$

Two types of capital

- physical capital

$$K_{t+1} = (1 - \delta) K_t + B_{t+1} G(I_t, K_t^*)$$

- intangible capital

$$K_{t+1}^* = [K_t^* - G(I_t, K_t^*)] (1 - \delta^*) + H(I_t^*, K_t)$$

- production of new physical capital

$$G(I_t, K_t^*) = \left[\phi (I_t)^{1-\psi} + (1 - \phi) (K_t^*)^{1-\psi} \right]^{\frac{1}{1-\psi}}$$

Representative household with recursive preferences

Technology modeled as permanent (growth rate) shocks

$$\log A_{t+1} - \log A_t = \Theta + Z_t + \Gamma W_{t+1}^1$$

where the 'long-run risk' component Z evolves as:

$$Z_{t+1} = \Psi Z_t + \Lambda W_{t+1}^2$$

Shock to **newly installed capital** is modeled as :

$$\log B_{t+1} = -\frac{1-\lambda}{\lambda} \left(Z_t + \Gamma W_{t+1}^1 \right)$$

Neutral technology growth rate and the investment specific shocks are perfectly negatively correlated.

The common component to $\log A_{t+1} - \log A_t$ and $\log B_{t+1}$ is $Z_t + \Gamma W_{t+1}^1$ where the 'long-run risk' component Z evolves as:

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Two shocks

- **immediate shock**: ΓW_{t+1}^1 (rescaled to have a unit sd)
- **long-run risk shock**: ΛW_{t+1}^2 (rescaled to have a unit sd)

Impact on $\log B_{t+1}$ temporary but impact on $\log A_{t+1}$ is permanent.

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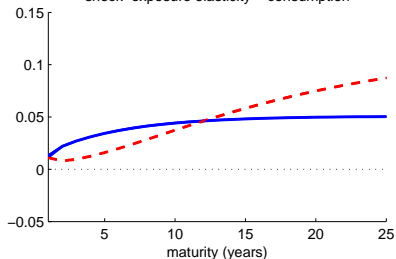
Questions

- What are the sources of risk premia? And what generates the heterogeneity?
 - Which shocks get priced (direct, long-run risk)?
 - At what horizons?
 - Through which channels (neutral, investment-specific)?

SHOCK ELASTICITIES FOR AGGREGATE CONSUMPTION

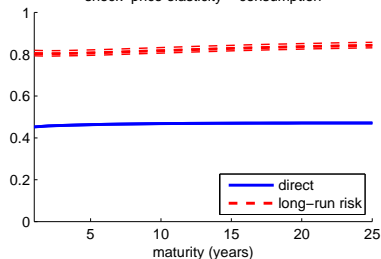
Shock-exposure elasticity

shock-exposure elasticity – consumption



Shock-price elasticity

shock-price elasticity – consumption

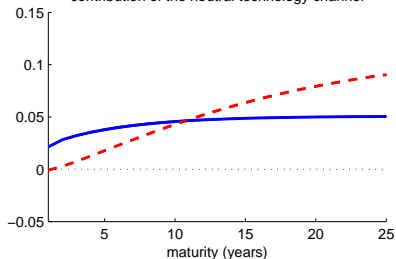


- shock-exposure elasticities gradually build (capital accumulation)
- shock-price elasticities are close to flat (recursive preferences)

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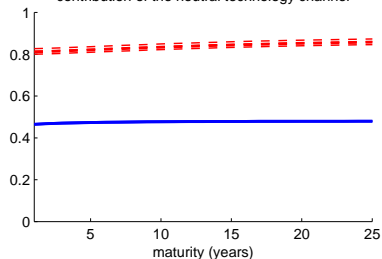
Shock-exposure elasticity

contribution of the neutral technology channel



Shock-price elasticity

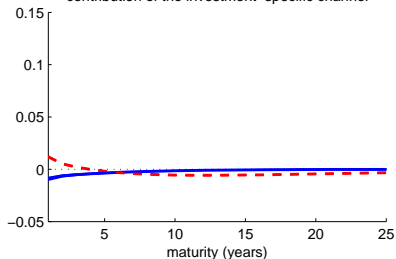
contribution of the neutral technology channel



- **shock-exposure elasticities** gradually build (capital accumulation)
- **shock-price elasticities** are close to flat (recursive preferences)
- **neutral technology growth rate shock** drives the pricing results

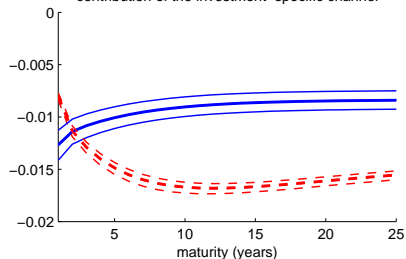
Shock-exposure elasticity

contribution of the investment-specific channel



Shock-price elasticity

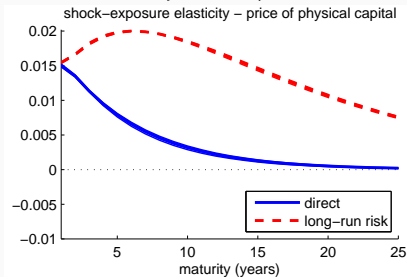
contribution of the investment-specific channel



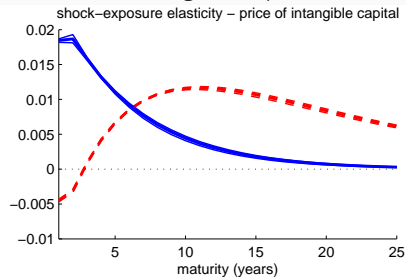
- **shock-exposure elasticities** gradually build (capital accumulation)
- **shock-price elasticities** are close to flat (recursive preferences)
- **neutral technology growth rate shock** drives the pricing results
- **shock to newly installed capital** has only small transitory impact on consumption dynamics \implies minimal impact on pricing

SHOCK ELASTICITIES FOR CAPITAL VALUATION

Physical capital

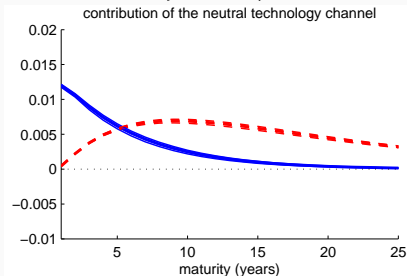


Intangible capital

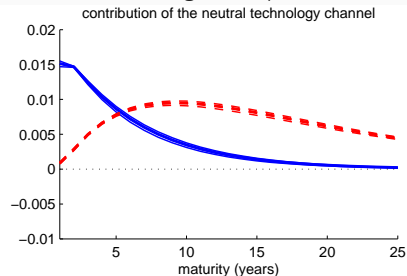


- dramatic difference in the responses to the long-run risk shock
- which channel contributes?

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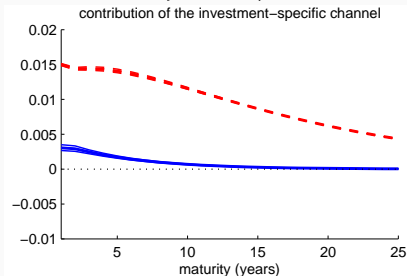


Intangible capital

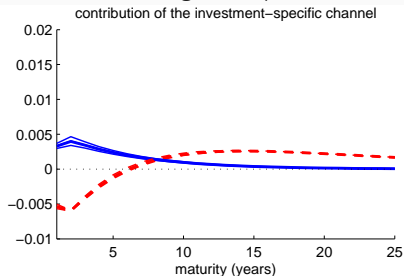


- dramatic difference in the responses to the long-run risk shock
- which channel contributes?
- responses propagated through the neutral technology channel are similar

Physical capital



Intangible capital



- dramatic difference in the responses to the long-run risk shock
- which channel contributes?
- responses propagated through the neutral technology channel are similar
- differences caused by the exposure to the long-run risk shock **propagated through the investment-specific channel**

Shock elasticities

- One-period pricing;
- Elasticities over multiple periods;
- Martingale components in stochastic discount factors

Applications

- Exponential-quadratic framework solvable in closed form;
 - Second-order perturbations of DSGE models (Dynare implementation);
 - Model of intangible capital [Ai, Croce and Li (2012)]
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A large class of models with frictions in the financial sector

- Households cannot invest in productive capital directly.
- Financing friction in banks casts a wedge between MRS and prices.
- Asset valuation depends on cash flows and ability to relax constraints.

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Gertler and Kiyotaki (2011)

- Households \iff financial sector \iff firms
- Occasionally binding constraint.
- Here: Series expansion around the deterministic steady state.
 - fits the exponential-quadratic framework.

- preferences

$$U_t(C, L) \doteq E \left[\sum_{j=0}^{\infty} \left(\log(C_{t+j} - hC_{t+j-1}) - \frac{\chi}{1+\varepsilon} (L_{t+j})^{1+\varepsilon} \right) \mid \mathcal{F}_t \right].$$

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- stochastic discount factor

$$\frac{S_{t+1}^h}{S_t^h} = \beta \frac{U_{c,t+1}}{U_{c,t}}$$

where U_c is the marginal utility of consumption

$$U_{c,t} = \frac{1}{C_t - hC_{t-1}} - h\beta E \left[\frac{1}{C_{t+1} - hC_t} \mid \mathcal{F}_t \right].$$

- preferences

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- priced asset: deposit contract

$$1 = E \left[\frac{S_{t+1}^h}{S_t^h} R_{t+1}^d \mid \mathcal{F}_t \right].$$

- objective

$$V(k_t, b_t, d_t) = \max_{(k_{t+1}, b_{t+1}, d_{t+1})} E \left[\frac{S_{t+1}^h}{S_t^h} [(1 - \sigma) n_{t+1} + \sigma V(k_{t+1}, b_{t+1}, d_{t+1})] \mid \mathcal{F}_t \right].$$

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- balance-sheet constraint

$$Q_t k_t = n_t + d_t + b_t$$

- accumulation of net worth

$$n_{t+1} = R_{t+1}^k Q_t k_t - R_{t+1}^b b_t - R_{t+1}^d d_t$$

- financial constraint

$$V(k_t, b_t, d_t) \geq \theta^k Q_t k_t - \theta^b b_t$$

- Neutral productivity shock

$$Y_t = A_t K_t^\alpha L_t^{1-\alpha}$$

- Capital-quality shock

$$K_{t+1} = \psi_{t+1} [I_t + (1 - \delta) K_t]$$

- Both modeled as stationary
 - but capital-quality shocks gain a lot of persistence through capital accumulation

- Return on capital

$$R_{t+1}^k = \psi_{t+1} \frac{Z_{t+1} + (1 - \delta) Q_{t+1}}{Q_t}$$

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- Revenue stream from a unit of capital produced at time t :

$$G_{t+j} = Z_{t+j} (1 - \delta)^{j-1} \prod_{m=1}^j \psi_{t+m}, \quad j \geq 1,$$

- corresponds to a multiplicative functional

$$\log G_{t+1} - \log G_t = \log(1 - \delta) + \log \psi_{t+1} + \log Z_{t+1} - \log Z_t.$$

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$$\log G_{t+1} - \log G_t = \log(1 - \delta) + \log \psi_{t+1} + \log Z_{t+1} - \log Z_t.$$

- Profit cash flow of the whole firm

$$G_t = \psi_t Z_t k_t$$

- Net worth paid to households

$$G_t = (1 - \sigma) n_t$$

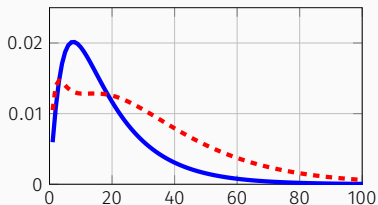
Euler equation

$$Q_t = E \left[\underbrace{\frac{S_{t+1}^h}{S_t^h} (1 - \sigma + \sigma \mu_{t+1}) \frac{1 + \lambda_t}{\mu_t + \lambda_t \theta^k}}_{\text{'stochastic discount factor'}} \underbrace{\psi_{t+1} (Z_{t+1} + (1 - \delta) Q_{t+1})}_{\text{payoff in period } t + 1} \mid \mathcal{F}_t \right]$$

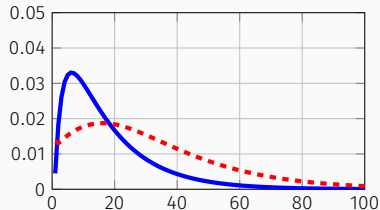
- λ_{t+1} Lagrange multiplier on the financial constraint
- μ_{t+1} Lagrange multiplier on the balance-sheet constraint
- frictionless economy: $\lambda_{t+1} = 0$, $\mu_{t+1} = 1$
- 'stochastic discount factor' is asset-specific (θ^k)

SHOCK-EXPOSURE ELASTICITIES

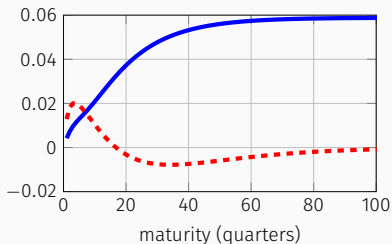
consumption shock-exposure



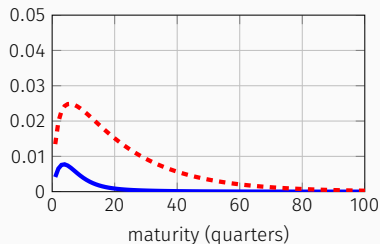
net-worth shock-exposure



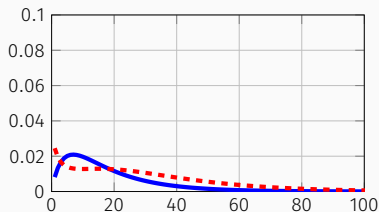
unit capital revenue shock-exposure



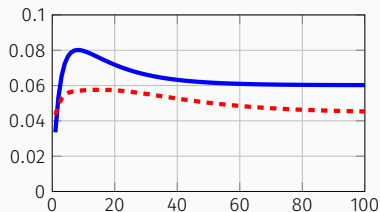
profit shock-exposure



household's stochastic discount factor



bank's stochastic discount factor for capital



$$\frac{S_{t+1}(\theta^k)}{S_t(\theta^k)} \doteq \frac{S_{t+1}^h}{S_t^h} (1 - \sigma + \sigma \mu_{t+1}) \frac{1 + \lambda_t}{\mu_t + \lambda_t \theta^k}$$

- shock-price elasticities for $S(\theta^k)$ are much larger ...
- ... and **do not decay to zero**.
- **state dependence (!!!)**

Shock elasticities

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Environment

- State dynamics

$$dX_t = \mu(X_t) dt + \sigma(X_t) dW_t$$

- Multiplicative processes (SDFs, cash flows)

$$d \log M_t = \beta(X_t) dt + \alpha(X_t) \cdot dW_t$$

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Shock elasticities

- Perturbation over a 'short time interval' dW_0
- Related to the Malliavian derivative

$$\varepsilon(x, t) = \nu(x) \cdot \left[\underbrace{\sigma(x)' \left(\frac{\partial}{\partial x} \log E \left[\frac{M_t}{M_0} \mid X_0 = x \right] \right)}_{\text{state propagation}} + \underbrace{\alpha(x)}_{\text{initial impact}} \right]$$

Central object in the analysis is the computation of

$$\phi_t(x) \doteq E \left[\frac{M_t}{M_0} \phi_0(X_t) \mid X_0 = x \right].$$

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Differential equation for $\phi_t(x)$

$$\frac{\partial}{\partial t} \phi_t = \left(\beta + \frac{1}{2} |\alpha|^2 \right) \phi_t + (\mu + \sigma \alpha) \cdot \frac{\partial \phi_t}{\partial x} + \frac{1}{2} \text{tr} \left[\sigma \sigma' \frac{\partial^2 \phi_t}{\partial x \partial x'} \right]$$

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Boundary conditions

- terminal: $\phi_0(x) \equiv 1$
- lateral: determined by the economics of the model

Solution

- finite differences, perturbation methods, MC simulations, ...

The Perron–Frobenius problem

$$E [M_t e(X_t) \mid X_0 = x] = \exp(\eta t) e(x)$$

The Perron–Frobenius problem

$$E [M_t e (X_t) | X_0 = x] = \exp (\eta t) e (x)$$

can be localized

$$\lim_{t \rightarrow 0} \frac{1}{t} [E [M_t e (X_t) | X_0 = x] - \exp (\eta) e (x)] = 0.$$

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Resulting PDE

$$\eta e = \left(\beta + \frac{1}{2} |\alpha|^2 \right) e + (\mu + \sigma \alpha) \cdot \frac{\partial e}{\partial x} + \frac{1}{2} \text{tr} \left[\sigma \sigma' \frac{\partial^2 e}{\partial x \partial x'} \right]$$

- Sturm–Liouville problem: finding $(\eta, e) \implies \phi_t(x) = \exp(\eta t) e(x)$

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· Sturm–Liouville problem: finding $(\eta, e) \implies \phi_t(x) = \exp(\eta t) e(x)$

Martingale factorization

$$M_t = \exp (\eta t) \tilde{M}_t \frac{e (X_0)}{e (X_t)}$$

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Applications

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Two types of agents: households and experts

- Households less efficient in managing capital
- Lending capital to experts is only possible using risk-free bond contracts
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Preferences

- Both agents linear preferences. Experts

$$E \left[\int_0^{\infty} e^{-\rho t} dC_t^h \mid \mathcal{F}_0 \right]$$

- Households more patient than experts: $r < \rho$.
- Marginal value of experts' wealth $\Theta_t \geq 1$ due to the financial constraint.

- wealth share of experts:

$$X_t = \frac{N_t}{Q_t K_t}$$

- aggregate capital

$$d \log K_t = \beta_k (X_t) dt + \bar{\alpha}_k dW_t$$

- drift depends on wealth distribution, since the two agents have different capital accumulation rates

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- drift depends on wealth distribution, since the two agents have different capital accumulation rates

State $X_t \in (0, \bar{x}]$

- X_t large — households invest all capital with firms
- X_t small — limited intermediation
- $X_t = \bar{x}$ — marginal value of experts' wealth declines to one, experts consume
 - reflecting boundary

Evolution of the state

$$dX_t = \mu(X_t) dt + \sigma(X_t) dW_t - X_t d\zeta_t$$

- $d\zeta_t$ – consumption of experts at $X_t = \bar{x}$.

Evolution of the state

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Stochastic discount factor of experts

$$\frac{S_t}{S_0} = e^{-\rho t} \frac{\theta(X_t)}{\theta(X_0)}$$

- $\Theta_t = \theta(X_t)$ – marginal utility of experts' wealth
 - not equal to one due to limited risk-sharing

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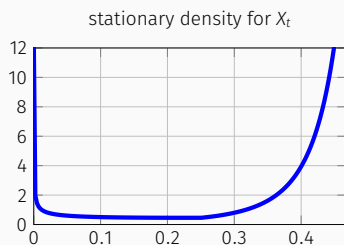
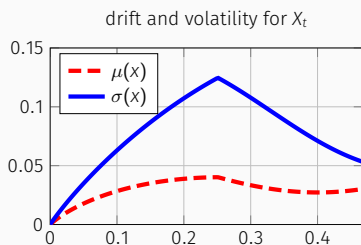
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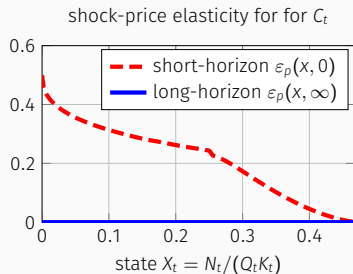
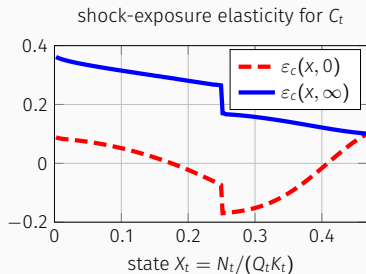
Cash flow: household consumption

$$d \log C_t = \beta_c(X_t) dt + \alpha_c(X_t) dW_t$$

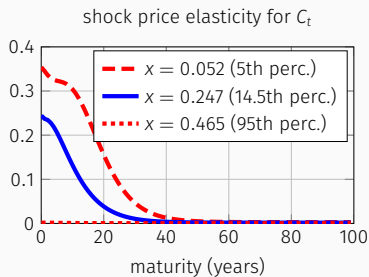
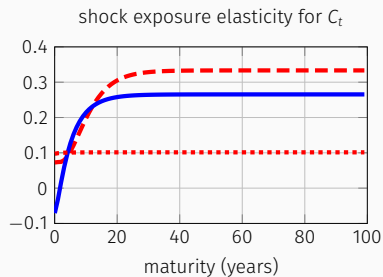


- Wealth in hands of experts valuable \implies accumulation, $\mu(x) > 0$
- Experts consume when $\theta(X_t)$ reaches one \implies reflecting right boundary
- Stationary density has two peaks, 'normal times' and 'financial crises'

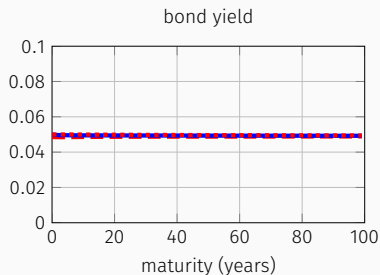
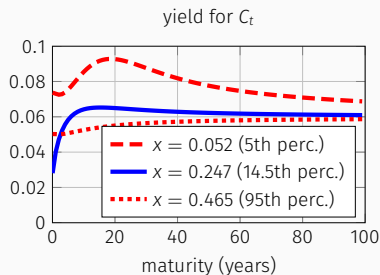
SHOCK ELASTICITIES IN THE SHORT AND LONG RUN



- Short-horizon consumption response small, even countercyclical.
- Shock-price elasticity is large during 'financial crises', zero in the long run.



- shock exposures generally increasing, shock prices decreasing



- Shock-exposure and shock-price elasticities explain the shape of yield differences between risky and risk-free cash flows
 - hump shaped in risky region
 - negative for short horizons in the center of the distribution

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Let Z_t be an n -state continuous time Markov chain

- realization a coordinate vector
- intensity matrix A

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- realization a coordinate vector
- intensity matrix A

Multiplicative functional

$$\log M_t = \sum_{0 \leq u \leq t} (Z_{u-})' \kappa Z_u + \int_0^t (Z_{u-})' \beta du + \int_0^t (Z_{u-})' \alpha dW_t$$

- β – elements are conditional (on Z_{u-}) growth rates
- α – rows are vectors of conditional exposures to dW_t
- κ – jump risk matrix, zeros on diagonal

Jump-risk perturbation

$$\log H_t(r) = \sum_{0 \leq u \leq t} (Z_{u-})' (r \kappa_d) Z_u + \int_0^t (Z_{u-})' \beta_h(r) du$$

- κ_d — perturbation direction (jump combination) that we are interested in

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- κ_d — perturbation direction (jump combination) that we are interested in

Shock elasticity function

$$\varepsilon(t) = \text{dvec} \left\{ \kappa_d (\hat{A} - A)' \right\} + \text{dvec} \left\{ [\Xi(t) \times \kappa_d] \hat{A}' \right\}$$

- \hat{A} — intensity matrix distorted by the martingale component of M
- $\Xi(t)$ traces out the transitory response

- Consumption modeled as multiplicative functional

$$\log C_t = \int_0^t (Z_{u-})' \beta_c du + \int_0^t (Z_{u-})' \alpha_c dW_t$$

- Agents endowed with Epstein-Zin preferences with $IES = 1$

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- Continuation value

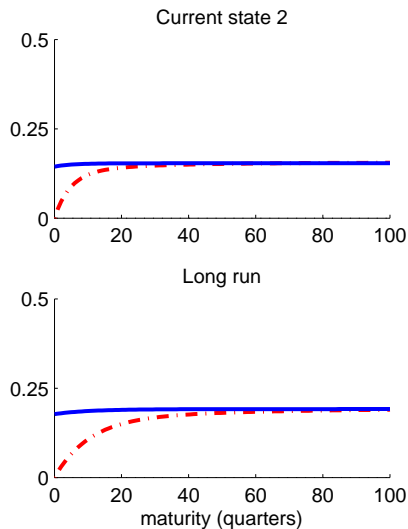
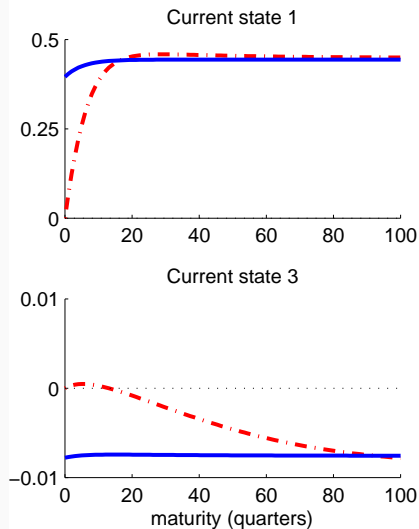
$$\log V_t = v \cdot Z_t + \log C_t$$

- Even though C_t does not jump at regime changes, V_t does:

$$\kappa_V = \mathbf{1}_n v' - v \mathbf{1}'_n$$

- Compare shock-price elasticities for CRRA and EZ stochastic discount factors

EXAMPLE: BONOMO AND GARCIA (1996)



• κ_d — normalized jump component of the martingale from $\log C$

Substantial information in the term structure of risk premia.

- yields on dividend strips and other maturity specific assets
 - empirical work by Kojien, van Binsbergen and others
 - many models produce counterfactual results
- holding period returns on long-term assets
 - entropy and variance bounds on the martingale component

Shock elasticities

- risk exposures and associated compensation
- term structure of risk

Theoretical framework

- continuous time
 - Borovička, Hansen, Hendricks, Scheinkman (2011); Borovička, Hansen (2016)
- discrete time
 - Borovička, Hansen (2014); Nakamura, Sergeyev, Steinsson (2016)
- non-Gaussian framework
 - Borovička, Hansen, Hendricks, Scheinkman (2011); Zviadadze (2016)
- nonlinear impulse response functions
 - Gallant, Rossi and Tauchen (1993); Koop, Pesaran and Potter (1996); Gouriéroux and Jasiak (2005); Borovička, Hansen, Scheinkman (2014)

Long-run pricing

- martingale decomposition
 - Hansen, Scheinkman (2009, 2012); Borovička, Hansen, Scheinkman (2016); Qin and Linetsky (2014a,b)
- role of the martingale component
 - Alvarez and Jermann (2005); Qin, Linetsky and Nie (2016); Bakshi, Chabi-Yo and Gao (2016); Christensen (2014); ...

Applications

- currency returns: Zviadadze (2016)
- long-run risk: Nakamura, Sergeyev, Steinsson (2016)

Empirical evidence

- downward sloping term structure of risk
 - van Binsbergen, Brandt and Koijen (2012); van Binsbergen, Hueskes, Koijen and Vrugt (2013); Ai, Croce, Diercks and Li (2013); Belo, Collin-Dufresne and Goldstein (2015); Hasler and Marfè (2015), Lopez, Lopez-Salido and Vazquez-Grande (2015), van Binsbergen and Koijen (2016)