



Macro, Money and Finance

Lecture 03: Macro-Finance Solution Technique

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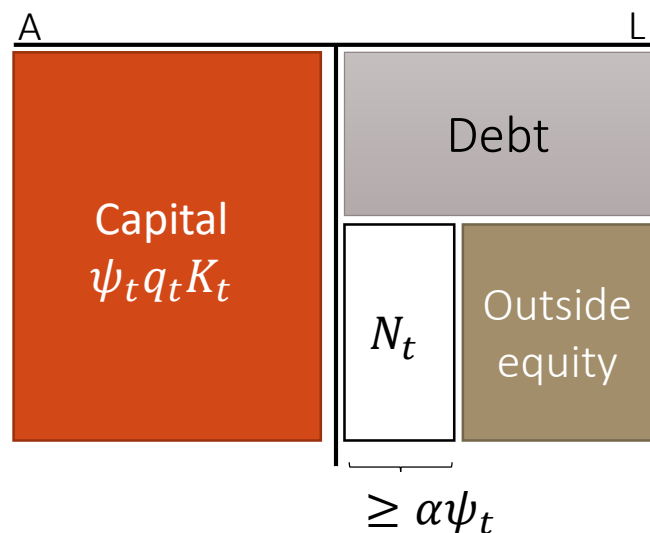
Desired Model Properties

- Normal regime: stable around steady state
 - Experts are adequately capitalized
 - Experts can absorb macro shock
- Endogenous risk
 - Fire-sales, liquidity spirals, fat tails
 - Spillovers across assets and agents
 - Market and funding liquidity connection
 - SDF vs. cash-flow news
- Volatility paradox
- Financial innovation less stable economy
- (“Net worth trap” double-humped stationary distribution)

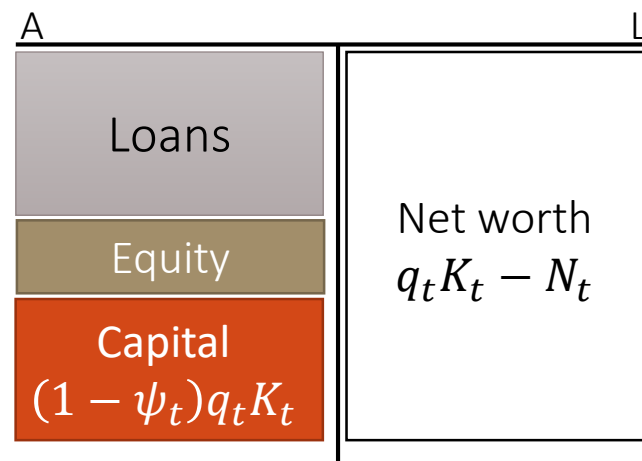
Two Type/Sector Model with Outside Equity

BruSan 2017: Handbook of Macroeconomics, Chapter 18, Section 3

Expert sector



Household sector



- Experts must hold fraction $\chi_t \geq \alpha \psi_t$ (skin in the game constraint)
- Return on inside equity N_t can differ from outside equity
 - Issue outside equity at required return from HH
 - In related model, He and Krishnamurthy 2013 impose that inside and outside equity have same return

Two Type Model Setup

Expert sector

▪ Output:

$$y_t = ak_t \quad a \geq \underline{a}$$

Household sector

▪ Output:

$$\underline{y}_t = \underline{a}k_t$$

$$A(\psi) = \psi a + (1 - \psi)\underline{a}$$

Capital share
of experts

Poll 4: Why is it important that households can hold capital?

- a) to capture fire-sales*
- b) for households to speculate*
- c) to obtain stationary distribution*

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Household sector

▪ Output:

$$\underline{y}_t = \underline{a}k_t$$

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Capital share
of experts

Poll 5: What are the modeling tricks to obtain stationary distribution?

a) switching types

b) agents die, OLG/perpetual youth models (without bequest motive)

c) different preference discount rates

Two Type Model Setup

Expert sector

- Output: $y_t = ak_t$ $a \geq \underline{a}$
- Consumption rate: c_t
- Investment rate: l_t

$$\frac{dk_t^i}{k_t^i} = (\Phi(l_t) - \delta)dt + \sigma dZ_t + \tilde{\sigma} d\tilde{Z}_t$$

capital evolution absent market transactions/fire-sales

Household sector

- Output: $y_t = \underline{a}k_t$
- Consumption rate: \underline{c}_t
- Investment rate: \underline{l}_t

$$\frac{d\underline{k}_t^i}{\underline{k}_t^i} = (\Phi(\underline{l}_t) - \delta)dt + \sigma dZ_t + \tilde{\sigma} d\tilde{Z}_t$$

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$$E_0 \left[\int_0^\infty e^{-\rho t} \frac{c_t^{1-\gamma}}{1-\gamma} dt \right]$$

Household sector

- Output: $\underline{y}_t = \underline{a}k_t$
- Consumption rate: \underline{c}_t
- Investment rate: \underline{l}_t

$$\frac{d\underline{k}_t^i}{\underline{k}_t^i} = (\Phi(\underline{l}_t) - \delta)dt + \sigma dZ_t + \tilde{\sigma} d\tilde{Z}_t$$

$$\rho \geq \underline{\rho} \quad E_0 \left[\int_0^\infty e^{-\underline{\rho} t} \frac{\underline{c}_t^{1-\gamma}}{1-\gamma} dt \right]$$

Two Type Model Setup

Expert sector

- Output: $y_t = ak_t$ $a \geq \underline{a}$

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Household sector

- Output: $y_t = \underline{a}k_t$

- Consumption rate: \underline{c}_t

- Investment rate: \underline{l}_t

$$\frac{d\underline{k}_t^i}{\underline{k}_t^i} = (\Phi(\underline{l}_t) - \delta)dt + \sigma dZ_t$$

- $E_0[\int_0^\infty e^{-\underline{\rho} t} \frac{\underline{c}_t^{1-\gamma}}{1-\gamma} dt]$ $\rho \geq \underline{\rho}$

Friction: Can only issue

- Risk-free debt

- Equity, but most hold $\chi_t \geq \alpha\psi_t$

|| Solving MacroModels Step-by-Step

0. Postulate aggregates, price processes & obtain return processes

1. For given SDF processes

static

- a. Real investment ι , (portfolio θ , & consumption choice of *each agent*)
 - *Toolbox 1*: Martingale Approach
- b. Asset/Risk Allocation *across types/sectors* & asset market clearing
 - *Toolbox 2*: “price-taking social planner approach” – Fisher separation theorem

2. Value functions

backward equation

- a. Value fcn. as fcn. of individual investment opportunities ω
 - *Special cases*
- b. De-scaled value fcn. as function of state variables η
 - *Digression*: HJB-approach (instead of martingale approach & envelop condition)
- c. Derive ζ price of risk, C/N -ratio from value fcn. envelop condition

3. Evolution of state variable η

forward equation

- *Toolbox 3*: Change in numeraire to total wealth (including SDF)
- (“Money evaluation equation” μ^ϑ)

4. Value function iteration & goods market clearing

- a. PDE of de-scaled value fcn.
- b. Value function iteration by solving PDE

0. Postulate Aggregates and processes

- Individual capital evolution:

$$\frac{dk_t^{\tilde{i},i}}{k_t^{\tilde{i},i}} = (\Phi(l^{\tilde{i},i}) - \delta)dt + \sigma dZ_t + d\Delta_t^{k,\tilde{i},i}$$

- Where $\Delta_t^{k,\tilde{i},i}$ is the individual cumulative capital purchase process

(c is numeraire)

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- Where $\Delta_t^{k,\tilde{l},i}$ is the individual cumulative capital purchase process
- Capital aggregation:

- Within sector i : $K_t^i \equiv \int k_t^{\tilde{l},i} d\tilde{l}$

- Across sectors: $K_t \equiv \sum_i K_t^i$

- Capital share: $\psi_t^i \equiv K_t^i / K_t$

$$\frac{dK_t}{K_t} = \int (\Phi(l^i) - \delta) di dt + \sigma dZ_t \quad \text{since } \Delta_t^{k,\tilde{l},i} \text{ add up to zero}$$

(c is numeraire)

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- Individual capital evolution:

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$$\frac{dK_t}{K_t} = \int (\Phi(l^i) - \delta) di dt + \sigma dZ_t$$

- Networth aggregation:

- Within sector i : $N_t^i \equiv \int n_t^{\tilde{l},i} d\tilde{l}$

- Across sectors: $\bar{N}_t \equiv \sum_i N_t^i$

- Wealth share: $\eta_t^i \equiv N_t^i / \bar{N}_t$

(c is numeraire)

0. Postulate Aggregates and processes

- Individual capital evolution:

$$\frac{dk_t^{\tilde{l},i}}{k_t^{\tilde{l},i}} = (\Phi(l^{\tilde{l},i}) - \delta)dt + \sigma dZ_t + d\Delta_t^{k,\tilde{l},i}$$

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- Value of capital stock: $q_t K_t$

Postulate $dq_t/q_t = \mu_t^q dt + \sigma_t^q dZ_t$

(c is numeraire)

0. Postulate Aggregates and processes

- Individual capital evolution:

$$\frac{dk_t^{\tilde{l},i}}{k_t^{\tilde{l},i}} = (\Phi(l^{\tilde{l},i}) - \delta)dt + \sigma dZ_t + d\Delta_t^{k,\tilde{l},i}$$

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$$\frac{dK_t}{K_t} = \int (\Phi(l^i) - \delta) di dt + \sigma dZ_t$$

Poll 14: How many Brownian motions span prob. space?

- Networth aggregation:

- Within sector i : $N_t^i \equiv \int n_t^{\tilde{l},i} d\tilde{l}$

- Across sectors: $\bar{N}_t \equiv \sum_i N_t^i$

- Wealth share: $\eta_t^i \equiv N_t^i / \bar{N}_t$

a) one

b) two

c) one + number of sectors

d) two + number of sectors

- Value of capital stock: $q_t K_t$

Postulate $dq_t/q_t = \mu_t^q dt + \sigma_t^q dZ_t$

(c is numeraire)

0. Postulate Aggregates and processes

- Individual capital evolution:

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Postulate $dq_t/q_t = \mu_t^q dt + \sigma_t^q dZ_t$

Same Brownian

(c is numeraire)

0. Postulate Aggregates and processes

- Individual capital evolution:

$$\frac{dk_t^{\tilde{i},i}}{k_t^{\tilde{i},i}} = (\Phi(l^{\tilde{i},i}) - \delta)dt + \sigma dZ_t + d\Delta_t^{k,\tilde{i},i}$$

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Postulate $dq_t/q_t = \mu_t^q dt + \sigma_t^q dZ_t$

- Postulated SDF-process: $\frac{d\xi_t^i}{\xi_t^i} = \underbrace{\mu_t^\xi}_{\equiv -r_t} + \underbrace{\sigma_t^{\xi^i}}_{\equiv -s_t^i} dZ_t$ (c is numeraire)

0. Postulate Aggregates and Processes

- ... from price processes to return processes (using Ito)
 - Use Ito product rule to obtain **capital gain rate** (in absence of purchases/sales)

- Define $\check{k}_t^{\tilde{i},i}$: $\frac{d\check{k}_t^{\tilde{i},i}}{\check{k}_t^{\tilde{i},i}} = (\Phi(l_t^{\tilde{i},i}) - \delta)dt + \sigma dZ_t + \cancel{d\Delta_t^{\tilde{k},\tilde{l},i}}$ without purchases/sales

$$dr_t^K(l_t^{\tilde{i},i}) = \left(\overbrace{\frac{A(\psi_t) - l_t^i}{q}}^{\text{Dividend yield}} + \overbrace{\Phi(l_t^i) - \delta + \mu_t^q + \sigma\sigma_t^q}^{\text{E[Capital gain rate]} = \frac{d(q_t \check{k}_t^i)}{(q_t \check{k}_t^i)}} \right) dt + (\sigma + \sigma_t^q) dZ_t$$

- Postulate SDF-process: (Example: $\xi_t^i = e^{-\rho t} u^{i'}(c_t)$)

$$\frac{d\xi_t^i}{\xi_t^i} = -r_t dt - \zeta_t^i dZ_t$$

↑
Price of risk

0. Postulate Aggregates and Processes

- ... from price processes to return processes (using Ito)
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- Postulate SDF-process: (Example: $\xi_t^i = e^{-\rho t} u^{i'}(c_t)$)

$$\frac{d\xi_t^i}{\xi_t^i} = -r_t dt - \zeta_t^i dZ_t$$

Poll 18: Why does drift of SDF equal risk-free rate

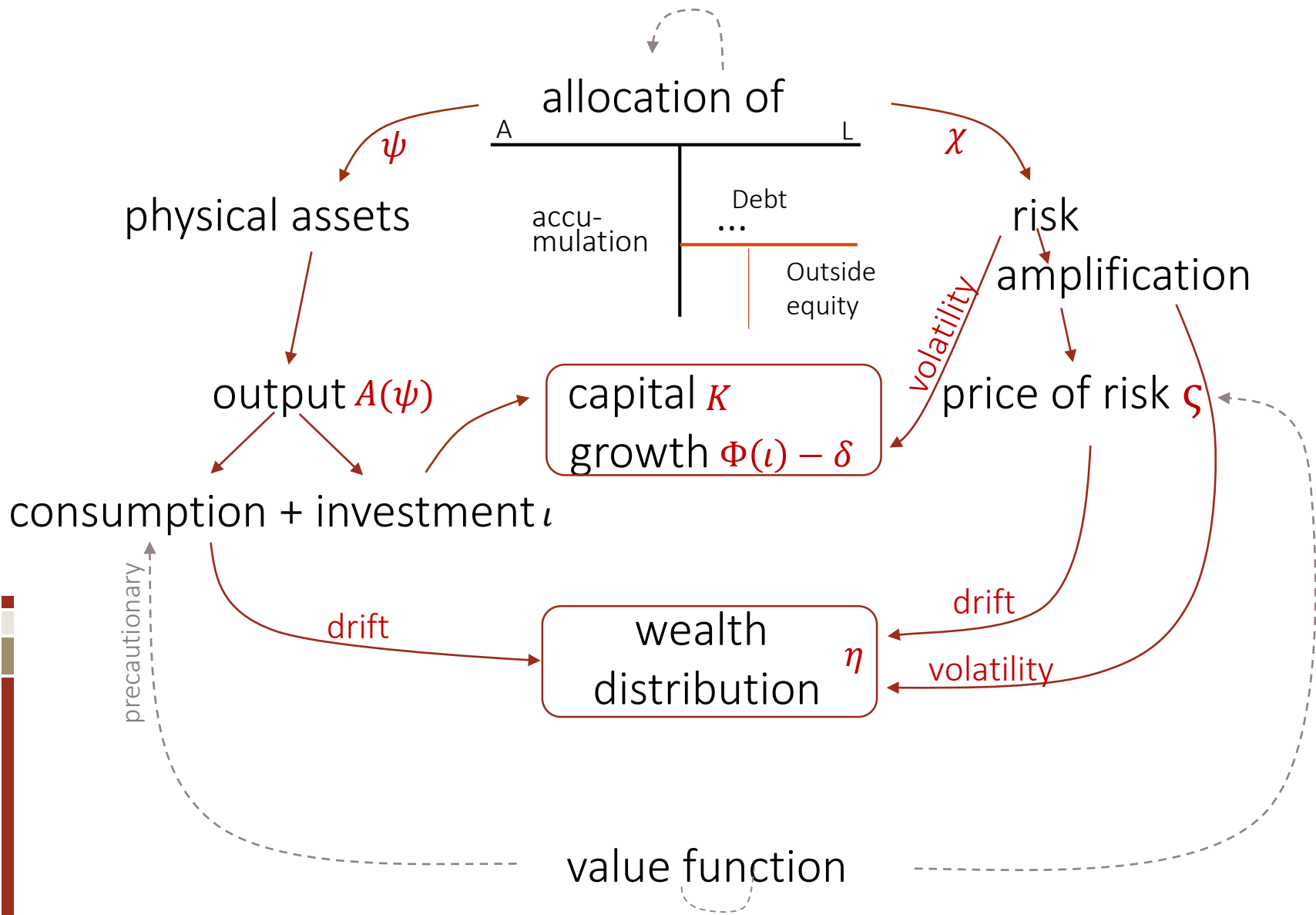
↑
Price of risk

a) no idio risk

b) $e^{-r^F} = E[SDF]^*1$

c) no jump in consumption

|| The Big Picture



Solving MacroModels Step-by-Step

0. Postulate aggregates, price processes & obtain return processes
1. For given SDF processes *static*
 - a. Real investment ι , (portfolio θ , & consumption choice of *each agent*)
 - *Toolbox 1*: Martingale Approach
 - b. Asset/Risk Allocation *across types/sectors* & asset market clearing
 - *Toolbox 2*: “price-taking social planner approach” – Fisher separation theorem
2. Value functions *backward equation*
 - a. Value fcn. as fcn. of individual investment opportunities ω
 - *Special cases*
 - b. De-scaled value fcn. as function of state variables η
 - *Digression*: HJB-approach (instead of martingale approach & envelop condition)
 - c. Derive ζ price of risk, C/N -ratio from value fcn. envelop condition
3. Evolution of state variable η *forward equation*
 - *Toolbox 3*: Change in numeraire to total wealth (including SDF)
 - (“Money evaluation equation” μ^ϑ)
4. Value function iteration & goods market clearing
 - a. PDE of de-scaled value fcn.
 - b. Value function iteration by solving PDE

1. Individual Agent Choice of ι , θ , c

- Choice of ι is static problem (and separable) for each t

- $$\max_{\iota^i} dr^K(\iota^i)$$
$$= \max_{\iota^i} \left(\frac{A(\psi) - \iota^i}{q} + \Phi(\iota^i) - \delta + \mu^q + \sigma\sigma^q \right)$$

- FOC: $\frac{1}{q} = \Phi'(\iota^i)$ Tobin's q

- All agents $\iota^i = \iota \Rightarrow \frac{dK_t}{K_t} = (\Phi(\iota) - \delta) dt + \sigma dZ_t$

- Special functional form:

- $\Phi(\iota) = \frac{1}{\kappa} \log(\kappa\iota + 1) \Rightarrow \kappa\iota = q - 1$

1a. Optimal Portfolio Choice

- Of experts with outside equity issuance (after plugging in households' outside equity choice)

$$\begin{aligned} \frac{a^{-l_t}}{q_t} + \Phi(l_t) - \delta + \mu_t^q + \sigma\sigma_t^q - r_t &= \\ &= [\zeta_t \chi_t / \psi_t + \underline{\zeta}_t (1 - \chi_t / \psi_t)] (\sigma + \sigma^q) \end{aligned}$$

- Of households' capital choice

$$\begin{aligned} \frac{\underline{a}^{-l_t}}{q_t} + \Phi(l_t) - \delta + \mu_t^q + \sigma\sigma_t^q - r_t &\leq \underline{\zeta}_t (\sigma + \sigma^q) \\ &\text{with equality if } \psi_t < 1 \end{aligned}$$

- New Approach replaces this step with Fisher Separation Social Planners' choice (see below)

1a. Individual Agent Choice of ι , θ , c

■ Consumption Choice: Martingale Approach

- Consider a self-financing trading strategy consisting of agent's networth *with consumption reinvested*.

$$\blacksquare \frac{d(\xi_t^i n_t)}{\xi_t^i n_t} + \frac{c_t}{n_t} dt = \underbrace{\left(-r_t + \mu_t^n - \varsigma_t^i \sigma_t^n + \frac{c_t}{n_t} \right)}_{=0} dt + \sigma \dots$$

$$\blacksquare \frac{c_t}{n_t} = r_t - \mu_t^n + \varsigma_t^i \sigma_t^n$$

- (only) useful for steady state characterization

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static

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 - *Toolbox 1*: Martingale Approach
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2. Value functions

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1b. Asset/Risk Allocation across I Types

■ Price-Taking Planner's Theorem:

A social planner that takes prices as given chooses a real asset (capital/production) allocation, ψ_t , and risk allocation χ_t , that coincides with the choices implied by all individuals' portfolio choices.

$$\begin{aligned}\boldsymbol{\varsigma}_t &= (\varsigma_t^1, \dots, \varsigma_t^I) \\ \boldsymbol{\chi}_t &= (\chi_t^1, \dots, \chi_t^I) \\ \boldsymbol{\sigma}(\boldsymbol{\chi}_t) &= (\sigma^1(\chi_t), \dots, \sigma^I(\chi_t))\end{aligned}$$

■ Planner's problem

$$\max_{\{\boldsymbol{\psi}_t, \boldsymbol{\chi}_t\}} E_t[dr_t^K(\boldsymbol{\psi}_t)] - \boldsymbol{\varsigma}_t \boldsymbol{\sigma}(\boldsymbol{\chi}_t) = dr^F \text{ in equilibrium}$$

subject to **friction**: $F(\boldsymbol{\psi}_t, \boldsymbol{\chi}_t) \leq 0$

■ Examples:

1. $\chi_t = \psi_t$ (if one holds capital, one has to hold risk)
2. $\chi_t \geq \alpha \psi_t$ (**skin in the game constraint**)

1b. Asset/Risk Allocation across I Types

■ Sketch of Proof of Theorem

1. Fisher Separation Thm: (delegated portfolio choice by firm)

- FOC yield the martingale approach solution
- Each individual agent (i, \tilde{i}) portfolio maximization is equivalent to the maximization problem of a firm

$$\max_{\{\theta^{j,i}\}} E_t \left[dr^{n(i,\tilde{i})} \right] - \zeta \sigma^{r^n}$$

- $dr^{n(i,\tilde{i})} = \sum_j \theta^{j,i} dr^j = \sum_j \theta^{j,i} E[dr^j] + \sum_j \theta^{j,i} \sigma^j dZ_t$

is linear in θ s

- Either bang-bang solution for θ s s.t. portfolio constraints bind
- Or prices/returns/risk premia are s.t. that firm is indifferent

2. Aggregate

- Taking η -weighted sum to obtain return on aggregate wealth

3. Use market clearing to relate θ s to ψ s & χ s (incl. θ -constraint)

1b. Allocation of Capital/Risk: 2 Types

- Expert: $\theta = (\theta^k, \theta^{oe}, \theta^d)$ for capital, outside equity, debt

A	L		
Physical capital θ^k	Debt	equity	O- equity

- Restrictions:
 - $\theta^k \geq 0,$
 - $\theta^{oe} \leq 0,$
 - $\theta^{oe} \geq -(1 - \alpha)\theta^k$

only issue outside equity
skin in the game

maximize

$$\theta^k E[dr^k] + \theta^{oe} E[dr^{oe}] + \theta^d r^f - \zeta_t (\theta^k + \theta^{oe}) \sigma^{r^k}$$

1b. Allocation of Capital/Risk: 2 Types

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only issue outside equity
skin in the game

maximize

$$\theta^k E[dr^k] + \theta^{oe} E[dr^{oe}] + \theta^d r^f - \zeta_t (\theta^k + \theta^{oe}) \sigma^{rk}$$

- Household: $\underline{\theta} = (\underline{\theta}^k, \underline{\theta}^{oe}, \underline{\theta}^d)$

maximize

$$\underline{\theta}^k E[d\underline{r}^k] + \underline{\theta}^{oe} E[d\underline{r}^{oe}] + \underline{\theta}^d r^f - \underline{\zeta}_t (\underline{\theta}^k + \underline{\theta}^{oe}) \sigma^{rk}$$

1b. Allocation of Capital/Risk: 2 Types

- Aggreate η -weighted sum of expert + HH max problem

$$\eta\{\dots\} + (1 - \eta)\{\dots\}$$

$$\begin{aligned} & \underbrace{\eta_t \theta_t^k}_{\psi_t :=} E[dr_t^k] + \underbrace{(1 - \eta_t) \underline{\theta}_t^k}_{=: \psi_t = 1 - \psi_t} E[d\underline{r}_t^k] + \\ & \underbrace{(\eta_t \theta_t^{oe} + (1 - \eta_t) \underline{\theta}_t^{oe})}_{=0} E[dr^{oe}] + \\ & \underbrace{(\eta_t \theta_t^d + (1 - \eta_t) \underline{\theta}_t^d)}_{=0} r_t^f \\ & - \underbrace{\varsigma_t \eta_t (\theta_t^k + \theta_t^{oe})}_{=: \chi_t} \sigma_t^{r^k} - \underline{\varsigma}_t \underbrace{(1 - \eta_t) (\underline{\theta}_t^k + \underline{\theta}_t^{oe})}_{=: \underline{\chi}_t} \sigma_t^{r^k} - \end{aligned}$$

1b. Allocation of Capital/Risk: 2 Types

- Aggreate η -weighted sum of expert + HH max problem

$$\eta\{\dots\} + (1 - \eta)\{\dots\}$$

$$\underbrace{\eta_t \theta_t^k}_{\psi_t :=} E[dr_t^k] + \underbrace{(1 - \eta_t) \underline{\theta}_t^k}_{=: \psi_t = 1 - \psi_t} E[d\underline{r}_t^k] +$$

Poll 33: Why = 0 ?

$$\underbrace{(\eta_t \theta_t^{oe} + (1 - \eta_t) \underline{\theta}_t^{oe})}_{=0} E[dr^{oe}] +$$

a) because marginal benefits = marginal costs at optimum

$$\underbrace{(\eta_t \theta_t^d + (1 - \eta_t) \underline{\theta}_t^d)}_{=0} r_t^f$$

b) due to martingale behavior

c) because outside equity and debt are in zero net supply

$$-\zeta_t \underbrace{\eta_t (\theta_t^k + \theta_t^{oe})}_{=: \chi_t} \sigma_t^{r^k} - \underline{\zeta}_t \underbrace{(1 - \eta_t) (\underline{\theta}_t^k + \underline{\theta}_t^{oe})}_{=: \underline{\chi}_t} \sigma_t^{r^k} -$$

1b. Allocation of Capital/Risk: 2 Types

- Translate constraints:

- $\chi_t \leq \psi_t$ experts cannot buy outside equity of others
only important for the case with idio risk

- $$\chi_t = \underbrace{\eta_t \theta_t^k}_{\psi_t} + \underbrace{\eta_t \theta_t^{oe}}_{\geq -\psi_t(1-\alpha)} \geq \alpha \psi_t$$

- Price-taking social planners problem

$$\max_{\{\psi_t, \underline{\psi}_t=1-\psi_t, \chi_t \in [\alpha\psi_t, \psi_t], \underline{\chi}_t=1-\chi_t\}} \frac{\psi_t a + \underline{\psi}_t \underline{a} - l_t}{q_t} + \Phi(l) - \delta - \zeta \chi_t \sigma_t^{rk} - \underline{\zeta}_t \underline{\chi}_t \sigma_t^{rk}$$

End of Proof. Q.E.D.

- Linear objective (if frictions take form of constraints)
 - Price of risk adjust such that objective becomes flat *or*
 - Bang-bang solution hitting constraints

1b. Allocation of Capital/Risk: 2 Types

- Example: 2 Types + no outside equity ($\alpha = 1$)

$$\max_{\{\psi_t, \chi_t\}} \frac{\psi_t a + (1 - \psi_t) \underline{a} - l_t}{q_t} - \chi_t \zeta_t (\sigma + \sigma_t^q) - (1 - \chi_t) \underline{\zeta}_t (\sigma + \sigma_t^q)$$

s.t. **friction** $\chi_t = \psi_t$ if no outside equity can be issued

- $FOC_{\chi}: \frac{a - \underline{a}}{q_t} = (\zeta_t - \underline{\zeta}_t) (\sigma + \sigma_t^q)$

- May hold only with inequality (\geq), if at constraint $\psi_t = 1$

1b. Allocation of Capital/Risk: 2 Types

- Example: 2 Type + with outside equity

$$\max_{\{\psi_t, \chi_t\}} \frac{\psi_t a + (1 - \psi_t) \underline{a} - \iota_t}{q_t} - \chi_t \zeta_t (\sigma + \sigma_t^q) - (1 - \chi_t) \underline{\zeta}_t (\sigma + \sigma_t^q)$$

- FOC_χ : Case 1: $\zeta_t (\sigma + \sigma_t^q) > \underline{\zeta}_t (\sigma + \sigma_t^q) \Rightarrow \chi_t = \alpha \psi_t$
 Case 2: $\qquad \qquad \qquad = \qquad \qquad \qquad \qquad \qquad \chi_t > \alpha \psi_t$

- Case 1: plug $\chi_t = \alpha \psi_t$ in objective

a. $FOC_\psi: \frac{a - \underline{a}}{q_t} \geq \alpha (\zeta_t - \underline{\zeta}_t) (\sigma + \sigma_t^q) \Rightarrow \psi_t = 1$

b. $\qquad \qquad \qquad = \qquad \qquad \qquad \qquad \qquad \Rightarrow \psi_t < 1$

- Case 2:

a. $FOC_\psi: \frac{a - \underline{a}}{q_t} > 0 \qquad \qquad \qquad \Rightarrow \psi_t = 1$

b. $\qquad \qquad \qquad = 0 \Rightarrow \psi_t < 1$ impossible

1b. Allocation of Capital/Risk: 2 Types

- Example: 2 Type + with outside equity

$$\max_{\{\psi_t, \chi_t\}} \frac{\psi_t a + (1 - \psi_t) \underline{a} - \iota_t}{q_t} - \chi_t \varsigma_t (\sigma + \sigma_t^q) - (1 - \chi_t) \underline{\varsigma}_t (\sigma + \sigma_t^q)$$

- FOC_χ : Case 1: $\varsigma_t (\sigma + \sigma_t^q) > \underline{\varsigma}_t (\sigma + \sigma_t^q) \Rightarrow \chi_t = \alpha \psi_t$

$$\text{Case 2:} \quad \quad \quad = \quad \quad \quad \chi_t > \alpha \psi_t$$

- Case 1: plug $\chi_t = \alpha \psi_t$ in objective

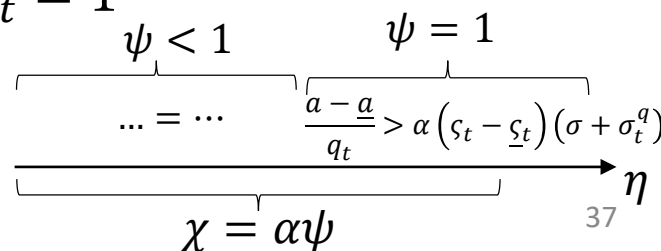
$$a. \quad FOC_\psi: \frac{a - \underline{a}}{q_t} \geq \alpha (\varsigma_t - \underline{\varsigma}_t) (\sigma + \sigma_t^q) \Rightarrow \psi_t = 1$$

$$b. \quad \quad \quad = \quad \quad \quad \Rightarrow \psi_t < 1$$

- Case 2:

$$a. \quad FOC_\psi: \frac{a - \underline{a}}{q_t} > 0 \quad \Rightarrow \psi_t = 1$$

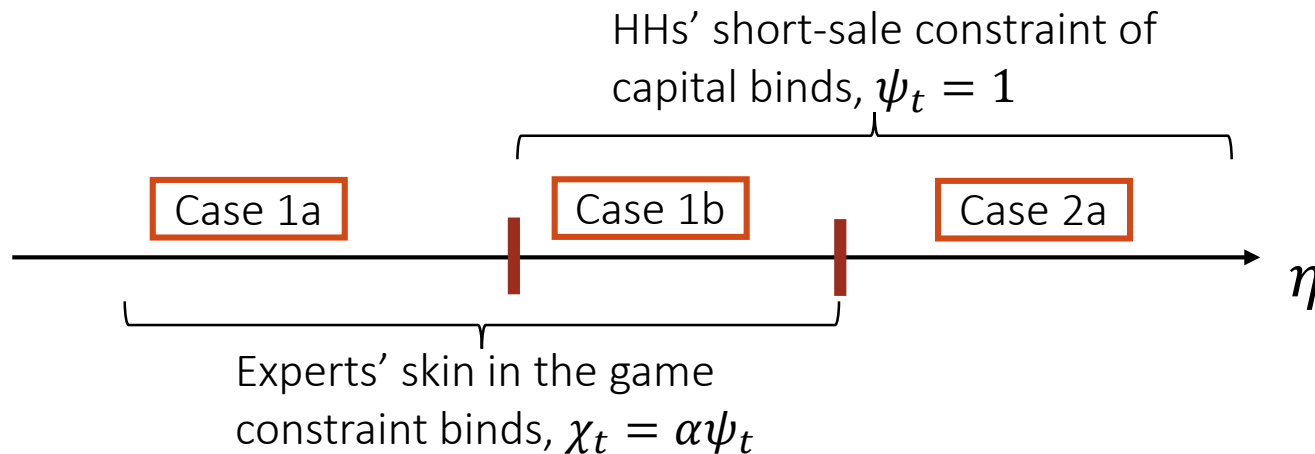
$$b. \quad \quad \quad = 0 \Rightarrow \psi_t < 1 \text{ impossible}$$



1b. Allocation of Capital, ψ , and Risk, χ

- Summarizing previous slide (2 types with outside equity)

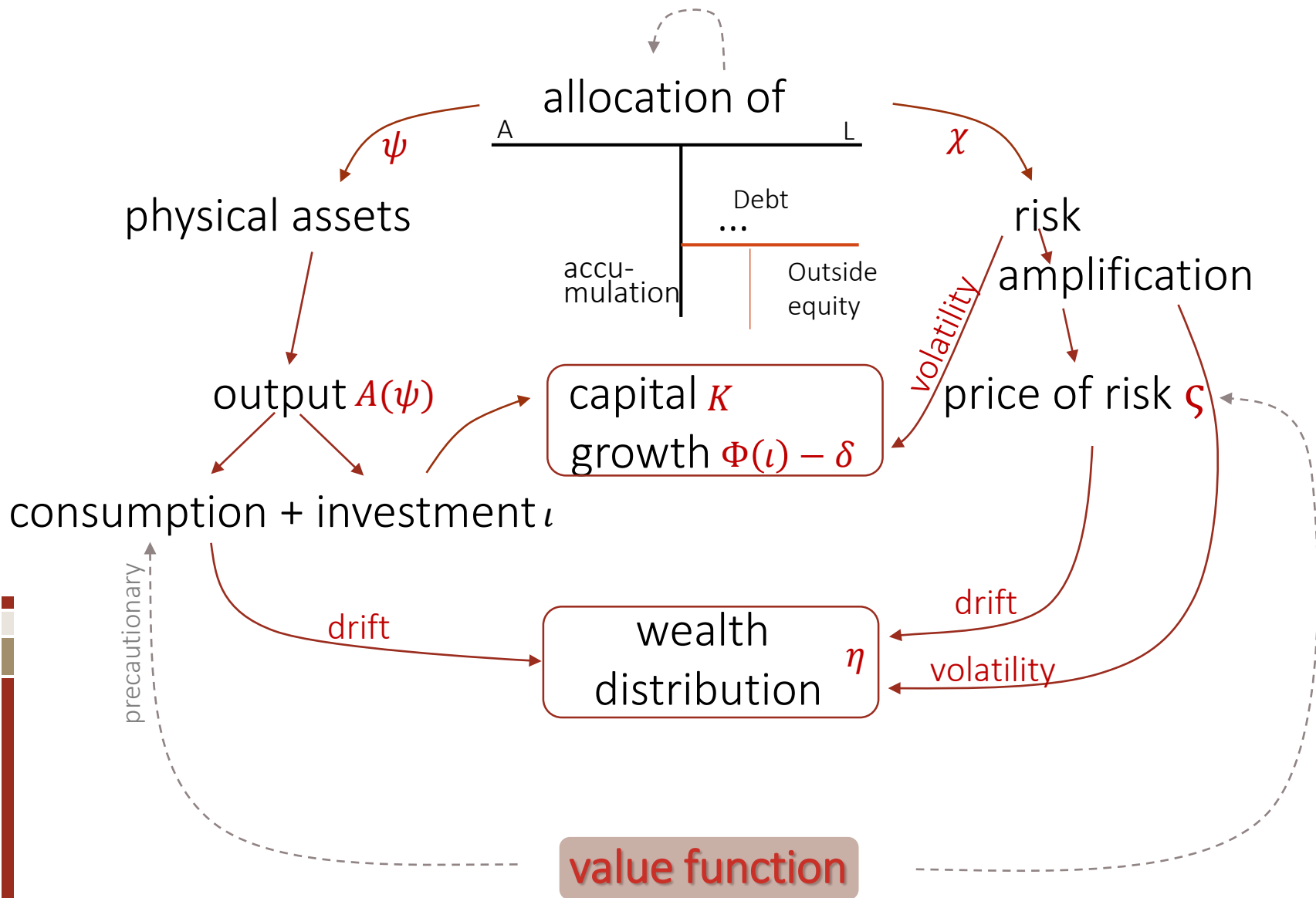
Cases	$\chi_t \geq \alpha\psi_t$	$\psi_t \leq 1$	$\frac{(a - \underline{a})}{q_t} \geq \alpha (\zeta_t - \underline{\zeta}_t) (\sigma + \sigma_t^q)$	$(\zeta_t - \underline{\zeta}_t) (\sigma + \sigma_t^q) \geq 0$
1a	=	<	=	>
1b	=	=	>	>
2a	>	=	>	=
impossible				



Solving MacroModels Step-by-Step

0. Postulate aggregates, price processes & obtain return processes
1. For given SDF processes *static*
 - a. Real investment ι , (portfolio θ , & consumption choice of *each agent*)
 - *Toolbox 1: Martingale Approach*
 - b. Asset/Risk Allocation *across types/sectors* & asset market clearing
 - *Toolbox 2: “price-taking social planner approach” – Fisher separation theorem*
2. Value functions *backward equation*
 - a. Value fcn. as fcn. of individual investment opportunities ω
 - *Special cases*
 - b. De-scaled value fcn. as function of state variables η
 - *Digression: HJB-approach (instead of martingale approach & envelop condition)*
 - c. Derive ζ price of risk, C/N -ratio from value fcn. envelop condition
3. Evolution of state variable η *forward equation*
 - *Toolbox 3: Change in numeraire to total wealth (including SDF)*
 - (“Money evaluation equation” μ^ϑ)
4. Value function iteration & goods market clearing
 - a. PDE of de-scaled value fcn.
 - b. Value function iteration by solving PDE

The Big Picture



2a. CRRA Value Function: relate to ω

Applies separately for each type of agent

- ω_t Investment opportunity/ “networth multiplier”

- CRRA/power utility $u(c) = \frac{c^{1-\gamma}-1}{1-\gamma}$

⇒ increase networth by factor, optimal consumption for all future states increases by same factor

⇒ $\left(\frac{c}{n}\right)$ -ratio is invariant in n

- ⇒ value function can be written as $\frac{u(\omega_t n_t)}{\rho}$, that is

$$= \frac{1}{\rho} \frac{(\omega_t n_t)^{1-\gamma} - 1}{1-\gamma} = \frac{1}{\rho} \frac{\omega_t^{1-\gamma} n_t^{1-\gamma} - 1}{1-\gamma}$$

- $\frac{\partial V}{\partial n} = \frac{\partial u}{\partial c}$ by optimal consumption (if no corner solution)

$$\frac{\omega_t^{1-\gamma} n_t^{-\gamma}}{\rho} = c_t^{-\gamma} \Leftrightarrow \frac{c_t}{n_t} = \rho^{1/\gamma} \omega_t^{1-1/\gamma}$$

Next step:

- Special simple cases
- replace ω_t with something scale invariant

2a. CRRA Value Function: Special Cases

$$\frac{c_t}{n_t} = \rho^{1/\gamma} \omega_t^{1-1/\gamma}$$

- For **log utility** $\gamma = 1$:

$$\xi_t = e^{-\rho t} / c_t = e^{-\rho t} / (\rho n_t) \text{ for any } \omega_t \Rightarrow \sigma_t^n = \sigma_t^c = \zeta_t$$

- Expected excess return: $\mu_t^A - r_t^F = \sigma_t^n \sigma_t^A$
- Recall $\frac{dn_t}{n_t} = -\frac{c_t}{n_t} dt + (1 - \theta) dr_t^K + \theta dr_t$

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- Recall $\frac{dn_t}{n_t} = -\frac{c_t}{n_t} dt + (1 - \theta) dr_t^K + \theta dr_t$

- For **constant investment opportunities** $\omega_t = \omega$,

$$\Rightarrow \frac{c}{n} \text{ is constant and hence } \sigma_t^c = \sigma^n$$

- Expected excess return: $\mu_t^A - r_t^F = \gamma \sigma_t^n \sigma_t^A$

Poll 43: Which term refers to (dynamic/Mertonian) hedging demand?

- γ
- σ_t^n
- hidden in risk-free rate
- none of the above

2a. CRRA Value Function: Special Cases

$$\frac{c_t}{n_t} = \rho^{1/\gamma} \omega_t^{1-1/\gamma}$$

- For **log utility** $\gamma = 1$:

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- Expected excess return: $\mu_t^A - r_t^F = \gamma \sigma_t^n \sigma_t^A$

$$\text{Now } \frac{dn_t}{n_t} = r^F dt + \frac{\zeta^2}{\gamma} dt + \frac{\zeta}{\gamma} dZ_t - \frac{c_t}{n_t} dt$$

$$r^F = \rho + \gamma \left(r^F + \frac{\zeta^2}{\gamma} - \frac{c_t}{n_t} \right) - \frac{\gamma+1}{2} \frac{\zeta^2}{\gamma} \qquad = \rho + \gamma \left(r^F - \frac{c_t}{n_t} \right) + \frac{\gamma-1}{\gamma} \frac{\zeta^2}{2}$$

$$\Rightarrow \frac{c_t}{n_t} = \rho + \frac{\gamma-1}{\gamma} \left(r^F - \rho + \frac{\zeta^2}{2\gamma} \right)$$

Also holds if ω_t evolves deterministically

2a. CRRA Value Function: Special Cases

$$\frac{c_t}{n_t} = \rho^{1/\gamma} \omega_t^{1-1/\gamma}$$

- For **log utility** $\gamma = 1$:

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- Recall $\frac{dn_t}{n_t} = -\frac{c_t}{n_t} dt + (1 - \theta) dr_t^K + \theta dr_t$

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2a. CRRA Value Function: Special Cases

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$$\Rightarrow \frac{c_t}{n_t} = \rho + \frac{\gamma-1}{\gamma} \left(r^F - \rho + \frac{\zeta^2}{2\gamma} \right)$$

Way to compute c_t/n_t if one can obtain from some other source r^F (omega can we avoided)

Also holds if ω_t evolves deterministically

2b. CRRA Value Fcn. & State Variable η

- Recall Martingale approach: if x_t is the value of a portfolio with return $\frac{dn_t}{n_t} + \frac{c_t}{n_t} dt$, then $\xi_t x_t$ must be a martingale

$$\frac{d(\xi_t n_t)}{\xi_t n_t} = -\frac{c_t}{n_t} dt + \text{martingale}$$

- Optimal consumption implies with CRRA- $V_t(n_t) = \frac{1}{\rho} \frac{(\omega_t n_t)^{1-\gamma}}{1-\gamma}$:

$$u'(c) = V'_t(n) \Leftrightarrow c_t^{-\gamma} = \frac{1}{\rho} \omega^{1-\gamma} n_t^{-\gamma} \Leftrightarrow e^{\rho t} \underbrace{e^{-\rho t} c_t^{-\gamma}}_{=\xi_t} n_t = \underbrace{\frac{1}{\rho} \omega^{1-\gamma} n_t^{1-\gamma}}_{(1-\gamma)V_t}$$

- Hence,

$$\frac{dV_t}{V_t} = \frac{d(e^{\rho t} \xi_t n_t)}{e^{\rho t} \xi_t n_t} = \left(\rho - \frac{c_t}{n_t} \right) dt + \text{martingale}$$

- Next, let's compute the drift of $\frac{dV_t}{V_t}$

2b. CRRA Value Fcn: De-scale by K_t

- Drift of $\frac{dV_t}{V_t}$, we could use Ito on $V_t = \frac{1}{\rho} \frac{(\omega_t n_t)^{1-\gamma}}{1-\gamma}$, but
 - *Poll 48: What could be the problem?*
 - a. Networth n_t is unbounded*
 - b. Networth $n_t(\eta_t)$ and N-multiplier $\omega_t(\eta_t)$ are not differentiable (if $q(\eta), p(\eta)$ have a kink).*
 - c. N-multiplier is not scale invariant*

2b. CRRA Value Fcn: De-scale by K_t

- Drift of $\frac{dV_t}{V_t}$, we could use Ito on $V_t = \frac{1}{\rho} \frac{(\omega_t n_t)^{1-\gamma}}{1-\gamma}$, but
 - Poll 49: What could be the problem?
 - a. Networth n_t is unbounded
 - b. Networth $n_t(\eta_t)$ and N-multiplier $\omega_t(\eta_t)$ are not differentiable (if $q(\eta), p(\eta)$ have a kink).
 - c. N-multiplier is not scale invariant
 - Answer: b.
- Let's de-scale the problem w.r.t. K_t

$$V_t = \frac{1}{\rho} \frac{(\omega_t n_t)^{1-\gamma}}{1-\gamma} = \underbrace{\frac{1}{\rho} \frac{\left(\omega_t \frac{n_t}{K_t}\right)^{1-\gamma}}{1-\gamma}}_{v_t :=} K_t^{1-\gamma}$$

and define v_t (which is twice differentiable in η_t)

- state variable K_t is easy to handle due to scale invariance

2b. CRRA Value Function

$$\frac{dV_t}{V_t} = \frac{d(v_t K_t^{1-\gamma})}{v_t K_t^{1-\gamma}}$$

- By Ito's product rule

$$= \left(\mu_t^v + (1-\gamma)(\Phi(\iota) - \delta) - \frac{1}{2}\gamma(1-\gamma)(\sigma^2) + (1-\gamma)\sigma\sigma_t^v \right) dt + \text{volatility terms}$$

- Recall by consumption optimality

$$\frac{dV_t}{V_t} - \rho dt + \frac{c_t}{n_t} dt \text{ follows a martingale}$$

- Hence, drift above = $\rho - \frac{c_t}{n_t}$

Still have to solve for μ_t^v, σ_t^v

Poll 51: Why martingale?

- Because we can "price" networth with SDF
- because ρ and c_t/n_t cancel out

2b. CRRA Value Fcn BSDE

- Only conceptual interim solution
 - We will transform it into a PDE in Step 4 below
- From last slide

$$\underbrace{\mu_t^v + (1 - \gamma)(\Phi(l) - \delta) - \frac{1}{2}\gamma(1 - \gamma)(\sigma^2) + (1 - \gamma)\sigma\sigma_t^v}_{=:\mu_t^V} = \rho - \frac{c_t}{n_t}$$

- Can solve for μ_t^v , then v_t must follow

$$\frac{dv_t}{v_t} = f(\eta_t, v_t, \sigma_t^v)dt + \sigma_t^v dZ_t$$

with

$$f(\eta_t, v_t, \sigma_t^v) = \rho - \frac{c_t}{n_t} - (1 - \gamma)(\Phi(l) - \delta) + \frac{1}{2}\gamma(1 - \gamma)(\sigma^2) - (1 - \gamma)\sigma\sigma_t^v$$

- Together with terminal condition v_T (possibly a constant for 1000 periods ahead), this is a **backward stochastic differential equation (BSDE)**
- A solution consists of processes v and σ^v
- Can use numerical BSDE solution methods (as random objects, so only get simulated paths)
- To solve this via a PDE we also need to get state evolution

||| Bellman Equation?

Poll 53: Where have we used the Bellman equation?

a) nowhere

b) it is hidden in the Martingale Approach

c) only needed in discrete time

Digression: HJB Approach

- Alternative to Martingale Approach
- Start from continuous-time analogue of Bellman equation

$$V(n, \omega) = \max_{\{c, \theta\}} E \left[\int_t^T e^{-\rho(s-t)} u(c_s) ds + e^{-\rho(T-t)} V(n_T, \omega_T) \mid n_t = n, \omega_t = \omega \right]$$

$$\text{s.t. } \frac{dn_t}{n_t} = -\frac{c_t}{n_t} dt + \sum_j \theta_t^j dr_t^j$$

- Subtract V

$$0 = \max_{\{c, \theta\}} E_t \left[\int_t^T e^{-\rho(s-t)} u(c_s) ds + \underbrace{e^{-\rho(T-t)} V(n_T, \omega_T) - e^{-\rho(t-t)} V(n_t, \omega_t)}_{= \int_t^T d(e^{-\rho(s-t)} V(n_s, \omega_s))} \right]$$

- Divide by $T - t$, take limit $T \rightarrow t$

$$\rho V(n_t, \omega_t) dt = \max_{\{c, \theta\}} u(c_t) dt + E_t [dV(n_t, \omega_t)]$$

s.t.

Digression: HJB Approach – the HJB Equation

- What is $E_t[dV(n_t, \omega_t)]$? If V is differentiable in n, ω

- Use Ito's Lemma

$$\frac{E_t[dV]}{dt} = \frac{\partial V}{\partial n} \mu^n n + \frac{\partial V}{\partial \omega} \mu^\omega \omega + \frac{1}{2} \frac{\partial^2 V}{\partial n^2} (\sigma^n n)^2 + \frac{1}{2} \frac{\partial^2 V}{\partial \omega^2} (\sigma^\omega \omega)^2 + \frac{\partial^2 V}{\partial \omega \partial n} \sigma^\omega \omega \sigma^n n$$

- Plug in μ^n, σ^n

- Hence HJB equation becomes

$$\rho V = \max_{c, \theta} \left(u(c) + \frac{\partial V}{\partial n} \left(-c + \sum_j \theta^j \mu^{r^j} n \right) + \frac{1}{2} \frac{\partial^2 V}{\partial n^2} \left(\sum_j \theta^j \sigma^{r^j} n \right)^2 + \right.$$

Digression: HJB Approach – the Value Fcn BSDE

- Two possibilities to get backward equation

1. Start from $\rho V(n_t, \omega_t) dt = u(c_t) dt + E_t[dV(n_t, \omega_t)]$

- Can write this as

$$\mu_t^V = \rho - \frac{u(c_t)}{V(n_t, \omega_t)}$$

- Plug in $V(n_t, \omega_t) = \frac{1}{\rho} \frac{(\omega_t n_t)^{1-\gamma}}{1-\gamma}$ and $\frac{c_t}{n_t} = \rho^{1/\gamma} \omega_t^{1-1/\gamma}$

$$\mu_t^V = \rho - \frac{c_t}{n_t}$$

- With $V = vK^{1-\gamma}$, get same BSDE for v as before

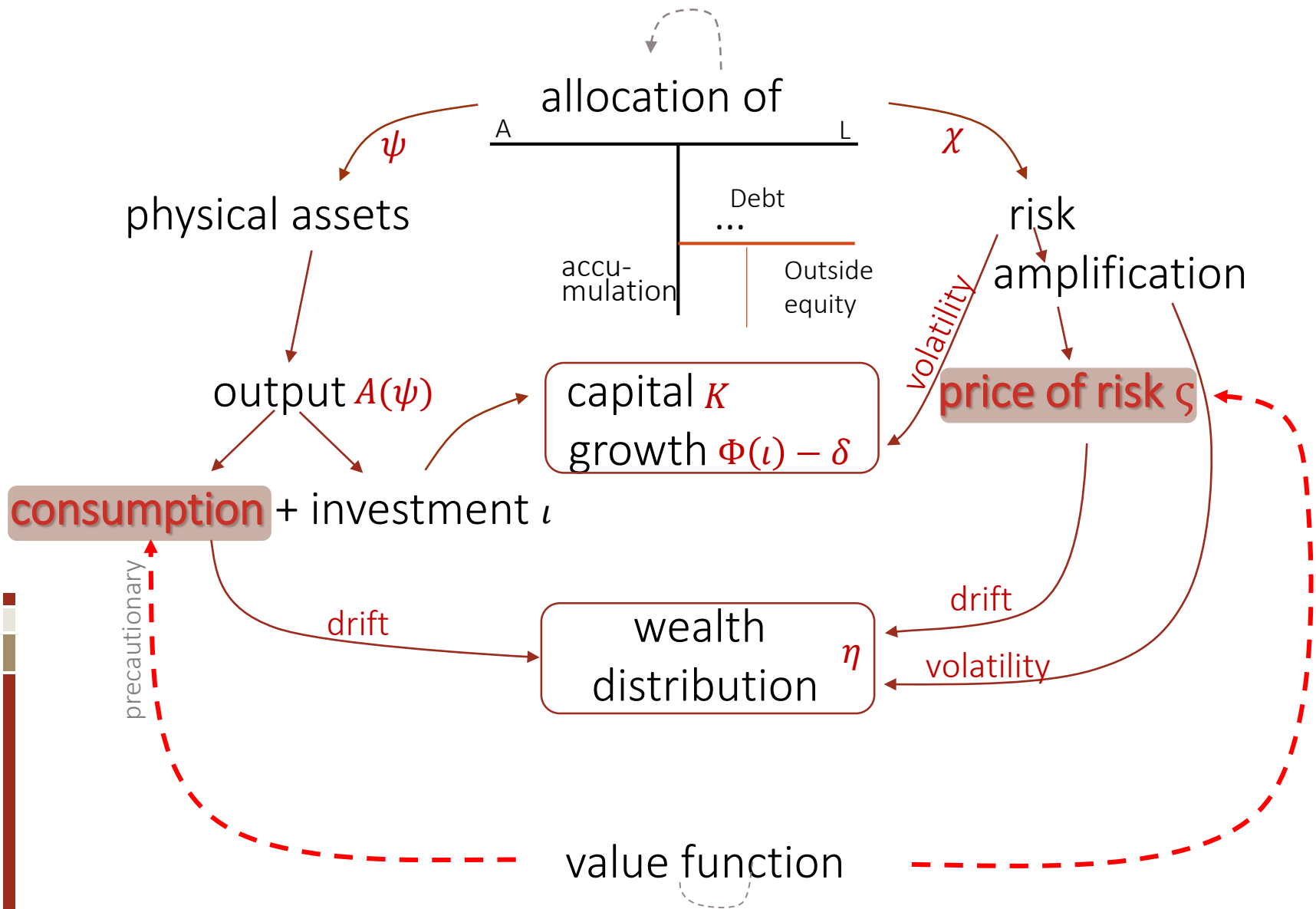
2. Use HJB equation

- Plug in $V(n_t, \omega_t) = \frac{1}{\rho} \frac{(\omega_t n_t)^{1-\gamma}}{1-\gamma}$ and its derivatives

$$\rho = \sum_j \theta^j \mu^{rj} - \frac{\gamma}{2} \left(\sum_j \theta^j \sigma^{rj} \right)^2 + (1-\gamma) \sigma^\omega \sum_j \theta^j \sigma^{rj} + \mu^\omega - \frac{\gamma}{2} (\sigma^\omega)^2$$

- This is a BSDE for ω (instead of v)

The Big Picture



2c. Get ζ s from Value Function Envelop

- Experts value function

$$v_t \frac{K_t^{1-\gamma}}{1-\gamma}$$

- To obtain $\frac{\partial V_t(n)}{\partial n_t}$ use $K_t = \frac{N_t}{\eta_t q_t} = \frac{n_t}{\eta_t q_t}$

$$V_t(n) = v_t \frac{n_t^{1-\gamma} / (\eta_t q_t)^{1-\gamma}}{1-\gamma}$$

- Envelop condition $\frac{\partial V_t(n)}{\partial n_t} = \frac{\partial u(c_t)}{\partial c_t}$

$$v_t \frac{n_t^{-\gamma}}{(\eta_t q_t)^{1-\gamma}} = c_t^{-\gamma}$$

- Using $K_t = \frac{n_t}{\eta_t q_t}$, $C_t = c_t$

$$\frac{v_t}{\eta_t q_t} K_t^{-\gamma} = C_t^{-\gamma}$$

$$\sigma_t^v - \sigma_t^\eta - \sigma_t^q - \gamma\sigma = -\gamma\sigma_t^c,$$

$$= -\zeta_t$$

- HH's value function

$$\underline{v}_t \frac{K_t^{1-\gamma}}{1-\gamma}$$

...

...

...

$$\sigma_t^{\underline{v}} - \sigma_t^{\underline{\eta}} - \sigma_t^{\underline{q}} - \underline{\gamma}\sigma = -\underline{\gamma}\sigma_t^{\underline{c}}$$

$$= -\underline{\zeta}_t$$

2c. Get $\frac{C_t}{N_t}$, $\frac{\underline{C}_t}{\underline{N}_t}$ from Value Function Envelop

■ Experts

Households

■ Recall $v_t \frac{n_t^{-\gamma}}{(\eta_t q_t)^{1-\gamma}} = c_t^{-\gamma}$

$$\frac{c_t}{n_t} = \frac{(\eta_t q_t)^{1/\gamma-1}}{v_t^{1/\gamma}}$$

$$\frac{C_t}{N_t} = \frac{(\eta_t q_t)^{1/\gamma-1}}{v_t^{1/\gamma}}$$

$$\frac{\underline{C}_t}{\underline{N}_t} = \frac{((1-\eta_t)q_t)^{1/\gamma-1}}{v_t^{1/\gamma}}$$

$$\frac{C_t + \underline{C}_t}{N_t + \underline{N}_t} = \eta_t \frac{C_t}{N_t} + (1 - \eta_t) \frac{\underline{C}_t}{\underline{N}_t}$$

Plug in from above

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3. GE: Markov States and Equilibria

- Equilibrium is a **map**

Histories of shocks $\{Z_\tau, 0 \leq \tau \leq t\}$ \dashrightarrow prices $q_t, \varsigma_t, \underline{\varsigma}_t, l_t, \theta_t, \psi_t, \chi_t$

wealth distribution

$$\eta_t = \frac{N_t}{q_t K_t} \in (0, 1)$$

wealth share

3. Law of Motion of Wealth Share η_t

- Method 1: Using Ito's quotation rule $\eta_t = N_t / (q_t K_t)$

- Recall

$$\frac{dN_t}{N_t} = r_t dt + \underbrace{\frac{\chi_t \psi_t (\sigma + \sigma_t^q)}{\eta_t}}_{\text{risk}} \underbrace{\zeta_t}_{\text{price of risk}} dt + \frac{\chi_t \psi_t (\sigma + \sigma_t^q)}{\eta_t} dZ_t - \frac{C_t}{N_t} dt$$

- $\frac{d\eta_t}{\eta_t} = \dots$ (lots of algebra)

- Method 2: Change of numeraire + Martingale Approach

- New numeraire: Total wealth in the economy, N_t
- Apply Martingale Approach for value of i 's portfolio
 - Simple algebra to obtain drift of η_t : μ_t^η
 Note that change of numeraire does not affect ratio η !

3. Aside: Change of Numeraire

- x_t^A is a value of a self-financing strategy/asset in \$
- Y_t price of € in \$ (exchange rate)

$$\frac{dY_t}{Y_t} = \mu_t^Y dt + \sigma_t^Y dZ_t$$

- x_t^A / Y_t value of the self-financing strategy/asset in €

$\underbrace{e^{-\rho t} u'(c_t)}_{=\xi_t} Y_t \frac{x_t^A}{Y_t}$ follows a martingale

$$\text{Recall } \mu_t^A - \mu_t^B = \underbrace{(-\sigma_t^\xi)}_{=\zeta_t} \underbrace{(\sigma^A - \sigma_t^B)}_{\text{risk}}$$

$$\mu_t^{A/Y} - \mu_t^{B/Y} = \underbrace{(-\sigma_t^\xi - \sigma_t^Y)}_{\text{price of risk}} \underbrace{(\sigma^A - \cancel{\sigma_t^Y} - \sigma_t^B + \cancel{\sigma_t^Y})}_{\text{risk}}$$

- Price of risk $\zeta^\epsilon = \zeta^\$ - \sigma^Y$

3. Aside: Change of Numeraire

- x_t^A is a value of a self-financing strategy/asset in \$
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- Price of risk $\zeta^\epsilon = \zeta^\$ - \sigma^Y$

Poll 64: Why does the price of risk change, though real risk remains the same

 - a) b/c risk-free rate might not stay risk-free*
 - b) b/c covariance structure changes*

3. μ^η Drift of Wealth Share: Many Types

- New Numeraire

- “Total wealth” in the economy, \overline{N}_t (without superscript)
 - Type i 's portfolio wealth = wealth share

- Martingale Approach with new numeraire

- Asset A = i 's portfolio return in terms of total wealth,

$$\left(\underbrace{\frac{C_t^i}{N_t^i}}_{\text{Dividend yield}} + \underbrace{\mu_t^{\eta^i}}_{\text{E[capital gains rate]}} \right) dt + \sigma_t^{\eta^i} dZ_t$$

- Asset B (benchmark asset that everyone can hold, e.g. risk-free asset or money (in terms of total economy wide wealth as numeraire))

$$r_t^M dt + \sigma_t^M dZ_t$$

- Apply our martingale asset pricing formula

$$\mu_t^A - \mu_t^B = \zeta_t^i (\sigma_t^A - \sigma_t^B)$$

3. μ^η Drift of Wealth Share: Many Types

- Asset pricing formula (relative to benchmark asset)

$$\mu_t^{\eta^i} + \frac{C_t^i}{N_t^i} - r_t^M = \left(\zeta_t^i - \sigma_t^{\bar{N}} \right) \left(\sigma_t^{\eta^i} - \sigma_t^M \right)$$

due to change

- Add up across types (weighted), in numeraire
(capital letters with bar & without superscripts are aggregates for total economy)

$$\underbrace{\sum_{i'} \eta_t^{i'} \mu_t^{\eta^{i'}}}_{=0} + \frac{\bar{C}_t}{\bar{N}_t} - r_t^M = \sum_{i'} \eta_t^{i'} \left(\zeta_t^{i'} - \sigma_t^{\bar{N}} \right) \left(\sigma_t^{\eta^{i'}} - \sigma_t^M \right)$$

Poll 66: Why = 0?

- Because we have stationary distribution
- Because η s sum up to 1
- Because η s follow martingale

3. μ^η Drift of Wealth Share: Many Types

- Asset pricing formula (relative to benchmark asset)

$$\mu_t^{\eta^i} + \frac{C_t^i}{N_t^i} - r_t^M = (\zeta_t^i - \sigma_t^{\bar{N}}) (\sigma_t^{\eta^i} - \sigma_t^M)$$

- Add up across types (weighted),
(capital letters with bar & without superscripts are aggregates for total economy)

$$\underbrace{\sum_{i'}^I \eta_t^{i'} \mu_t^{\eta^{i'}}}_{=0} + \frac{\bar{C}_t}{\bar{N}_t} - r_t^M = \sum_{i'} \eta_t^{i'} (\zeta_t^{i'} - \sigma_t^{\bar{N}}) (\sigma_t^{\eta^{i'}} - \sigma_t^M)$$

- Subtract from each other yield **wealth share dynamics**

$$\begin{aligned} \mu_t^{\eta^i} &= (\zeta^i - \sigma^N) (\sigma^{\eta^i} - \sigma^M) - \sum_{i'} \eta_t^{i'} (\zeta_t^{i'} - \sigma_t^N) (\sigma_t^{\eta^{i'}} - \sigma_t^M) \\ &\quad - \left(\frac{C_t^i}{N_t^i} - \frac{\bar{C}_t}{\bar{N}_t} \right) \end{aligned}$$

3. μ^η Drift of Wealth Share: Two Types

- Asset pricing formula (relative to benchmark asset)

$$\mu_t^\eta + \frac{C_t}{N_t} - r_t^M = (\zeta - \sigma_t^{\bar{N}}) (\sigma^\eta - \sigma^M)$$

- Add up across types (weighted),
(capital letters without superscripts are aggregates for total economy)

$$\underbrace{(\eta_t \mu_t^\eta + (1 - \eta_t) \mu_t^\eta)}_{=0} + \frac{C_t}{N_t} - r_t^M =$$

$$\eta_t (\zeta_t - \sigma_t^{\bar{N}}) (\sigma_t^\eta - \sigma_t^M) - (1 - \eta_t) (\zeta_t - \sigma_t^{\bar{N}}) (\sigma_t^\eta - \sigma_t^M)$$

- Subtract from each other yields **wealth share drift**

$$\mu_t^\eta = (1 - \eta_t) (\zeta_t - \sigma_t^{\bar{N}}) (\sigma_t^\eta - \sigma_t^M) - (1 - \eta_t) (\zeta_t - \sigma_t^{\bar{N}}) (\sigma_t^\eta - \sigma_t^M)$$

$$- \left(\frac{C_t}{N_t} - \frac{C_t + \underline{C}_t}{q_t K_t} \right)$$

3. σ^η Volatility of Wealth Share

- Since $\eta_t^i = N_t^i / \bar{N}_t$,

$$\begin{aligned} \sigma_t^{\eta^i} &= \sigma_t^{N^i} - \sigma_t^{\bar{N}} = \sigma_t^{N^i} - \sum_{i'} \eta_t^{i'} \sigma_t^{N^{i'}} \\ &= (1 - \eta_t^i) \sigma_t^{N^i} - \sum_{i^- \neq i} \eta_t^{i^-} \sigma_t^{N^{i^-}} \end{aligned}$$

- Note for 2 types example

Change in notation in 2 type setting
Type-networth is $n = N^i$

$$\sigma_t^\eta = (1 - \eta_t) (\sigma_t^n - \sigma_t^q)$$

$$\sigma_t^n = \underbrace{\chi_t / \eta_t}_{=\theta^k + \theta^{oe}} (\sigma + \sigma_t^q)$$

$$\sigma_t^q = \frac{1 - \chi_t}{1 - \eta_t} (\sigma + \sigma_t^q)$$

- Hence,

$$\sigma_t^\eta = \frac{\chi_t - \eta_t}{\eta_t} (\sigma + \sigma_t^q)$$

- Note also, $\eta_t \sigma_t^\eta + (1 - \eta_t) \sigma_t^q = 0 \Rightarrow \sigma_t^\eta = -\frac{\eta_t}{1 - \eta_t} \sigma_t^q$

3. Amplification Formula: Loss Spiral

- Recall $\sigma_t^\eta = \underbrace{\frac{\chi_t - \eta_t}{\eta_t}}_{\text{leverage}} (\sigma + \sigma_t^q)$
- By Ito's Lemma on $q(\eta)$ $\sigma_t^q = \frac{q'(\eta_t)}{q(\eta_t)} \eta_t \sigma_t^\eta$

$$\sigma_t^q = \underbrace{\frac{q'(\eta_t)}{q/\eta_t}}_{\text{elasticity}} \frac{\chi_t - \eta_t}{\eta_t} (\sigma + \sigma_t^q)$$

- Total volatility

$$\sigma + \sigma_t^q = \frac{\sigma}{1 - \frac{q'(\eta_t)\chi_t - \eta_t}{q/\eta_t}}$$

- Loss spiral

- Market illiquidity (price impact elasticity)

3. Amplification Formula: Loss Spiral

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$$\sigma_t^q = \underbrace{\frac{q'(\eta_t)}{q/\eta_t}}_{\text{elasticity}} \frac{\chi_t - \eta_t}{\eta_t} (\sigma + \sigma_t^q)$$

- Total volatility

$$\sigma + \sigma_t^q = \frac{\sigma}{1 - \frac{q'(\eta_t)\chi_t - \eta_t}{q/\eta_t}}$$

Poll 71: Where is the spiral?

- a) Sum of infinite geometric series (denominator)
- b) in q' , since with constant price, no spiral

- Loss spiral

- Market illiquidity (price impact elasticity)

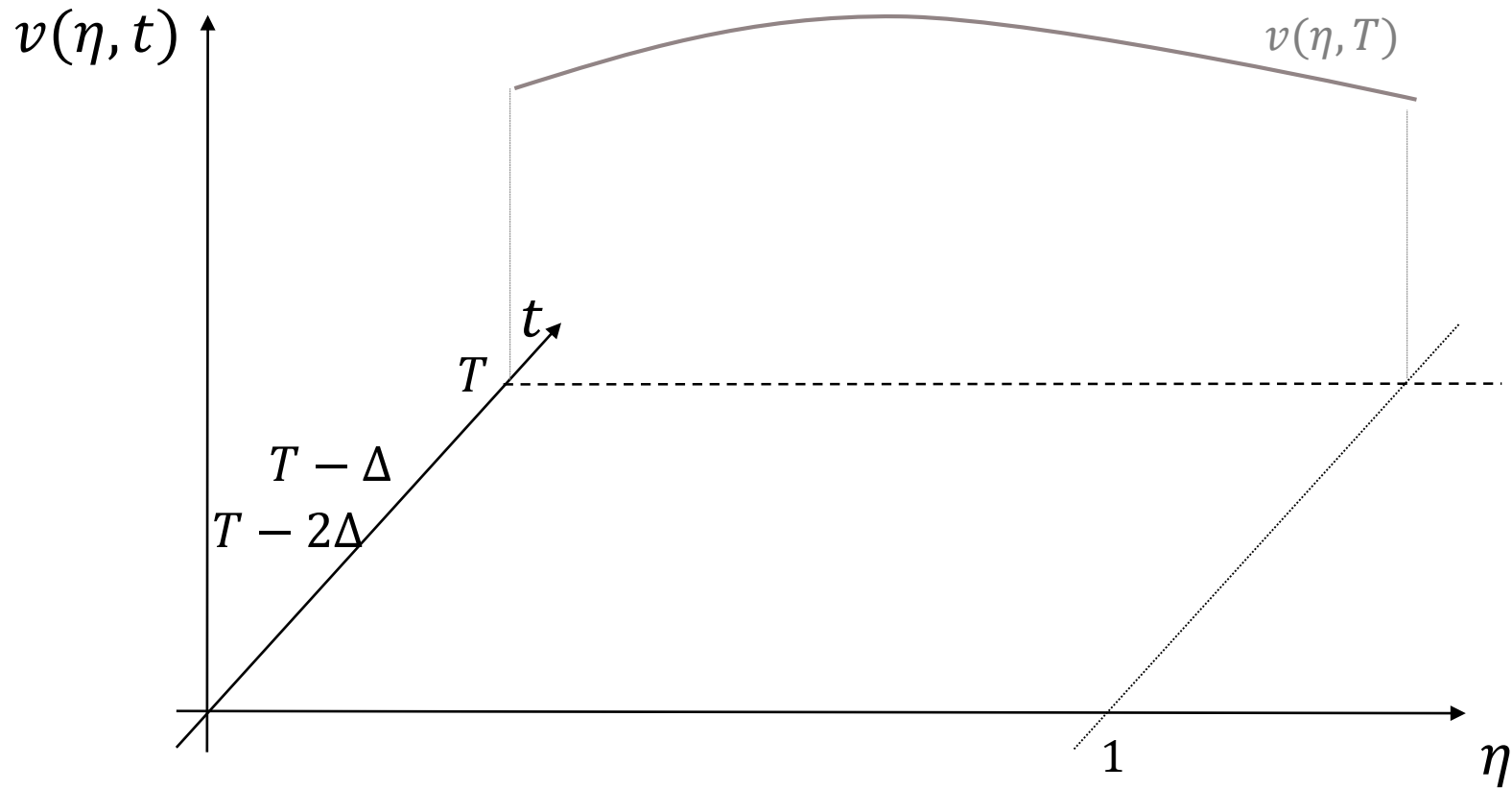
|| Solving MacroModels Step-by-Step

0. Postulate aggregates, price processes & obtain return processes
1. For given SDF processes *static*
 - a. Real investment ι , (portfolio θ , & consumption choice of *each agent*)
 - *Toolbox 1*: Martingale Approach
 - b. Asset/Risk Allocation *across types/sectors* & asset market clearing
 - *Toolbox 2*: “price-taking social planner approach” – Fisher separation theorem
2. Value functions *backward equation*
 - a. Value fcn. as fcn. of individual investment opportunities ω
 - *Special cases*
 - b. De-scaled value fcn. as function of state variables η
 - *Digression*: HJB-approach (instead of martingale approach & envelop condition)
 - c. Derive ζ price of risk, C/N -ratio from value fcn. envelop condition
3. Evolution of state variable η *forward equation*
 - *Toolbox 3*: Change in numeraire to total wealth (including SDF)
 - (“Money evaluation equation” μ^ϑ)
4. Value function iteration & goods market clearing
 - a. PDE of de-scaled value fcn.
 - b. Value function iteration by solving PDE

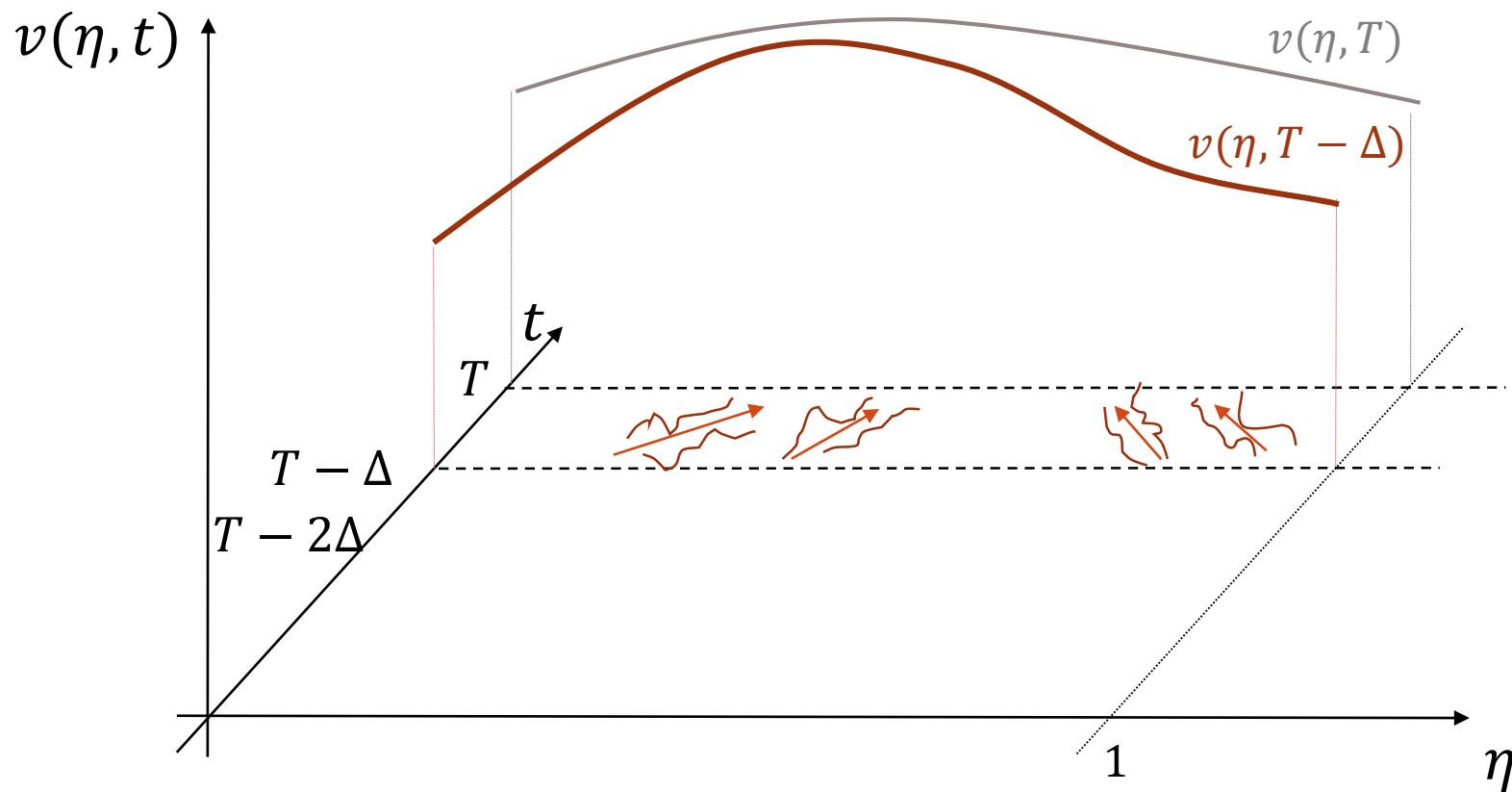
4. Value function Iteration - Big picture

- Add time, t , as an additional state variable $v(\eta, t)$, $\underline{v}(\eta, t)$
- Convert BSDE into PDE

4. Value Function Iteration – Big Picture



4. Value Function Iteration – Big Picture



4. Value function Iteration - Big picture

- Add time, t , as an additional state variable $v(\eta, t)$, $\underline{v}(\eta, t)$

- Convert BSDE into PDE using Ito's Lemma

Short-hand notation:
 $\partial_x f$ for $\partial f / \partial x$

- $\mu_t^v v_t = \partial_t v_t + \eta_t \mu_t^\eta \partial_\eta v_t + \frac{1}{2} (\eta_t \sigma_t^\eta)^2 \partial_{\eta\eta} v_t$

- $\mu_t^{\underline{v}} \underline{v}_t = \partial_t \underline{v}_t + \eta_t \mu_t^\eta \partial_\eta \underline{v}_t + \frac{1}{2} (\eta_t \sigma_t^\eta)^2 \partial_{\eta\eta} \underline{v}_t$

- Propose any arbitrary value function $v(\eta, T)$ and $\underline{v}(\eta, T)$ (far in the future $t = T$)

- ... and iterate back to $t = 0$

- In each step use

- From Step 2: $\mu_t^v v_t, \mu_t^{\underline{v}} \underline{v}_t$

- From Step 3: $\eta_t \mu_t^\eta$ and $\eta_t \sigma_t^\eta$ (η -evolution)

- Portfolio choice, planners' problem, (static conditions)

- Market clearing

- To calculate all terms in these $\mu_{t-\Delta}^v v_{t-\Delta}, \eta_{t-\Delta} \mu_{t-\Delta}^\eta$ and $\eta_{t-\Delta} \sigma_{t-\Delta}^\eta$

4a. PDE Expert Value Function Iteration

- Postulate $v_t = v(\eta_t, t)$

Short-hand notation:
 $\partial_x f$ for $\partial f / \partial x$

- By Ito's Lemma

- $$\frac{dv_t}{v_t} = \underbrace{\frac{\partial_t v_t + \partial_\eta v_t \eta \mu_t^\eta + \frac{1}{2} \partial_{\eta\eta} v_t (\eta_t \sigma_t^\eta)^2}{v_t}}_{\mu_t^v} dt + \underbrace{\frac{\partial_\eta v_t \eta \sigma_t^\eta}{v_t}}_{\sigma_t^v} dZ_t$$

- That is,

- $$\mu_t^v v_t = \partial_t v_t + \partial_\eta v_t \eta \mu_t^\eta + \frac{1}{2} \partial_{\eta\eta} v_t (\eta_t \sigma_t^\eta)^2$$

- $$\sigma_t^v v_t = \partial_\eta v_t \eta \sigma_t^\eta$$

- Equating with Step 2 \Rightarrow "growth equation"

$$\begin{aligned} & \partial_t v_t + (\eta \mu_t^\eta + (1 - \gamma) \sigma \eta_t \sigma_t^\eta) \partial_\eta v_t + \frac{1}{2} \partial_{\eta\eta} v_t (\eta_t \sigma_t^\eta)^2 = \\ & = \left(\rho - (1 - \gamma)(\Phi(\iota) - \delta) + \frac{1}{2} \gamma(1 - \gamma)(\sigma^2) \right) v_t - \frac{c_t}{n_t} v_t \end{aligned}$$

4a. PDE Expert Value Fcn: Replacing Terms

$$\begin{aligned} & \partial_t v_t + (\eta_t \mu_t^\eta + (1 - \gamma) \sigma \eta_t \sigma_t^\eta) \partial_\eta v_t + \frac{1}{2} \partial_{\eta\eta} v_t (\eta_t \sigma_t^\eta)^2 = \\ & = \left(\rho - (1 - \gamma)(\Phi(\underline{l}) - \delta) + \frac{1}{2} \gamma (1 - \gamma) (\sigma^2) \right) v_t - \frac{C_t}{n_t} v_t \end{aligned}$$

1. Replace "blue terms" using results from Step 3.

$$\begin{aligned} \mu_t^\eta &= (1 - \eta_t) (\underline{s}_t - \sigma_t^q - \sigma) \left(\sigma_t^\eta - \underbrace{\sigma_t^M}_{=0} \right) \\ &\quad - (1 - \eta_t) \left(\underline{s}_t - \sigma_t^q - \sigma \right) \left(\sigma_t^\eta - \underbrace{\sigma_t^M}_{=0} \right) - \left(\frac{C_t}{N_t} - \frac{C_t + \underline{C}_t}{q_t K_t} \right) \\ \sigma_t^\eta &= \frac{\chi_t^{-\eta_t}}{\eta_t} (\sigma + \sigma_t^q) & \sigma_t^\eta &= -\frac{\eta_t}{1 - \eta_t} \sigma_t^\eta \end{aligned}$$

2. Replace "tanned terms" using results from Step 2c.

$$\begin{aligned} \underline{s}_t &= -\sigma_t^v + \sigma_t^\eta + \sigma_t^q + \gamma \sigma, & \underline{s}_t &= -\sigma_t^v + \sigma_t^\eta + \sigma_t^q + \underline{\gamma} \sigma \\ \frac{C_t}{N_t} &= \frac{(\eta_t q_t)^{1/\gamma - 1}}{v_t^{1/\gamma}} & \frac{\underline{C}_t}{\underline{N}_t} &= \frac{((1 - \eta_t) q_t)^{1/\underline{\gamma} - 1}}{v_t^{1/\underline{\gamma}}} \end{aligned}$$

Recall from Ito's Lemma
 $\sigma_t^v v_t = \partial_\eta v_t \eta_t \sigma_t^\eta$

3. Replace "red terms" $\underline{l}_t, \sigma_t^q, \chi_t, \underline{\chi}_t$ (see below)

4a. PDE HH Value Fcn: Replacing Terms

$$\begin{aligned} \partial_t \underline{v}_t + (\eta_t \mu_t^\eta + (1 - \gamma) \sigma \eta_t \sigma_t^\eta) \partial_\eta \underline{v}_t + \frac{1}{2} \partial_{\eta\eta} \underline{v}_t (\eta_t \sigma_t^\eta)^2 &= \\ = \left(\rho - (1 - \gamma) (\Phi(l) - \delta) + \frac{1}{2} \gamma (1 - \gamma) (\sigma^2) \right) \underline{v}_t - \frac{C_t}{N_t} \underline{v}_t \end{aligned}$$

1. Replace “blue terms” using results from Step 3.

$$\mu_t^\eta = \dots$$

$$\sigma_t^\eta = \frac{\chi_t^{-\eta_t}}{\eta_t} (\sigma + \sigma_t^q)$$

$$\sigma_t^\eta = -\frac{\eta_t}{1 - \eta_t} \sigma_t^\eta$$

2. Replace “tanned terms” using results from Step 2c.

$$\zeta_t = -\sigma_t^v + \sigma_t^\eta + \sigma_t^q + \gamma \sigma,$$

$$\frac{C_t}{N_t} = \frac{(\eta_t q_t)^{1/\gamma - 1}}{v_t^{1/\gamma}}$$

Recall from Ito's Lemma
 $\sigma_t^v v_t = \partial_\eta v_t \eta \sigma_t^\eta$

$$\underline{\zeta}_t = -\sigma_t^v + \sigma_t^\eta + \sigma_t^q + \underline{\gamma} \sigma$$

$$\frac{\underline{C}_t}{\underline{N}_t} = \frac{((1 - \eta_t) q_t)^{1/\underline{\gamma} - 1}}{\underline{v}_t^{1/\underline{\gamma}}}$$

3. Replace “red terms” $l_t, \sigma_t^q, \chi_t, \underline{\chi}_t$ (see below)

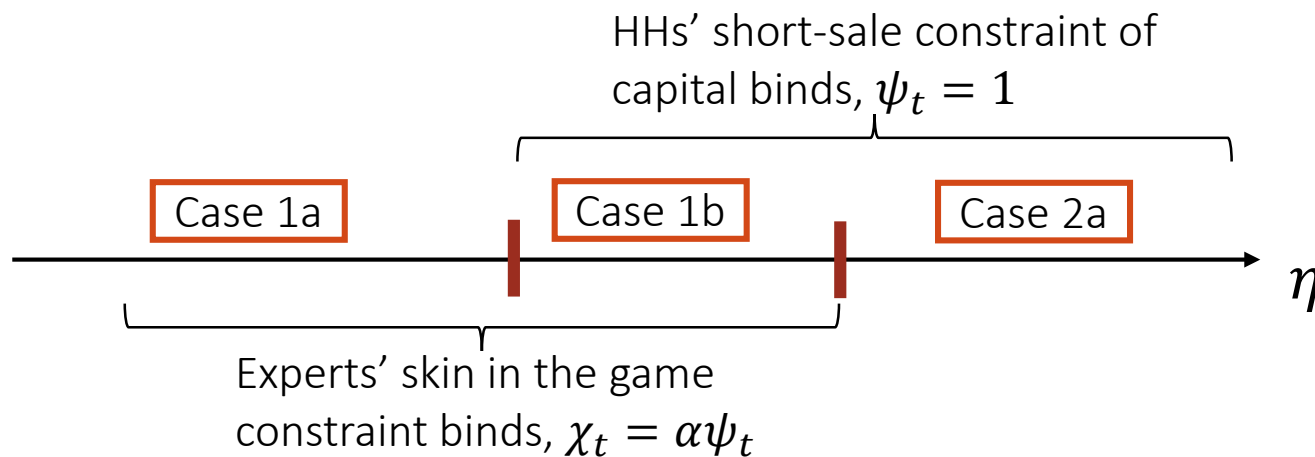
4a. Replacing l_t

- Recall from optimal re-investment $\Phi'(l_t) = 1/q_t$
 - For $\Phi(l) = \frac{1}{\kappa} \log(\kappa l + 1) \Rightarrow \kappa l = q - 1$

4a. Replacing χ , obtain ψ for good mkt clearing

- Recall from planner's problem (Step 1b)

Cases	$\chi_t \geq \alpha\psi_t$	$\psi_t \leq 1$	$\frac{(a - \underline{a})}{q_t} \geq \alpha (\zeta_t - \underline{\zeta}_t) (\sigma + \sigma_t^q)$	$(\zeta_t - \underline{\zeta}_t) (\sigma + \sigma_t^q) \geq 0$
1a	=	<	=	>
1b	=	=	>	>
2a	>	=	>	=



4a. Replacing χ , obtain ψ for good mkt clearing

- Need to determine diff in risk premia $(\zeta_t - \underline{\zeta}_t) (\sigma + \sigma_t^q)$:

- Recall

- diff in price of risk:
$$\zeta_t - \underline{\zeta}_t = -\sigma_t^v + \sigma_t^{\underline{v}} + \frac{\sigma_t^\eta}{1-\eta_t}$$

- By Ito's lemma
$$\sigma_t^v = \frac{v'}{v} \eta_t \sigma_t^\eta \text{ and } \sigma_t^{\underline{v}} = \frac{v'}{\underline{v}} \eta_t \sigma_t^\eta$$

$$\Rightarrow (\zeta_t - \underline{\zeta}_t) (\sigma + \sigma_t^q) = \left(-\frac{v'}{v} + \frac{v'}{\underline{v}} + \frac{1}{(1-\eta_t)\eta_t} \right) \eta_t \sigma_t^\eta (\sigma + \sigma_t^q)$$

$$= \left(-\frac{v'}{v} + \frac{v'}{\underline{v}} + \frac{1}{(1-\eta_t)\eta_t} \right) (\chi_t - \eta_t) (\sigma + \sigma_t^q)^2$$

- Note, since $-\frac{v'}{v} + \frac{v'}{\underline{v}} + \frac{1}{(1-\eta_t)\eta_t} > 0$,

$$(\zeta_t - \underline{\zeta}_t) (\sigma + \sigma_t^q) > 0 \Leftrightarrow \chi_t > \eta_t \Leftrightarrow \alpha > \eta_t$$

4a. Replacing χ , obtain ψ for good mkt clearing

- Determination of ψ_t

$$(a - \underline{a})/q_t \geq \underline{\alpha} \left(-\frac{v'}{v} + \frac{\underline{v}'}{\underline{v}} + \frac{1}{(1 - \eta_t)\eta_t} \right) (\chi_t - \eta_t)(\sigma + \sigma_t^q)^2$$

with equality if $\psi_t < 1$

- Determination of χ_t

$$\chi_t = \max\{\alpha\psi_t, \eta_t\}$$

4a. Market Clearing

- Output good market

$$C_t = (\psi_t a + (1 - \psi_t)\underline{a} - \iota_t)K_t$$

- ... jointly restricts ψ_t and q_t

$$\psi_t a + (1 - \psi_t)\underline{a} - \iota(q) = \underbrace{\left(\frac{\eta_t q_t}{v_t}\right)^{1/\gamma}}_{C_t/K_t} + \underbrace{\left(\frac{(1 - \eta_t)q_t}{\underline{v}_t}\right)^{1/\underline{\gamma}}}_{\underline{C}_t/K_t}$$

4a. Market Clearing

- Output good market

$$C_t = (\psi_t a + (1 - \psi_t)\underline{a} - \iota_t)K_t$$

- Capital market is taken care off by price taking social planner approach

$$1 - \theta_t = \frac{\psi_t q_t K_t}{\eta_t q_t K_t}$$

- Risk-free debt also taken care off by price taking social planner approach (would be cleared by Walras Law anyways)

4a. $\sigma^q(q, q')$

- Recall from “amplification slide” – Step 3

$$\sigma + \sigma_t^q = \frac{\sigma}{1 - \frac{q'(\eta_t)}{q/\eta_t} \frac{\chi_t - \eta_t}{\eta_t}}$$

$$\sigma^q = \frac{q'(\eta_t)}{q(\eta_t)} (\chi_t - \eta_t) (\sigma + \sigma_t^q)$$

- Now all red terms are replaced and we can solve ...

4b. Algorithm – Static Step

- Suppose we know functions $v(\eta), \underline{v}(\eta)$, have five static conditions:

- $\kappa l_t = q_t - 1$

- Planner condition for ψ_t

- Planner condition for χ_t

- $$\psi_t a + (1 - \psi_t) \underline{a} - \iota(q) = \underbrace{\left(\frac{\eta_t q_t}{v_t}\right)^{1/\gamma}}_{C_t/K_t} + \underbrace{\left(\frac{(1-\eta_t)q_t}{\underline{v}_t}\right)^{1/\gamma}}_{\underline{C}_t/K_t}$$

- $$\sigma^q = \frac{q'(\eta_t)}{q(\eta_t)} (\chi_t - \eta_t) (\sigma + \sigma_t^q)$$

⇒ Get
 $q(\eta)$,
 $\psi(\eta)$,
 $\sigma^\eta(\eta)$

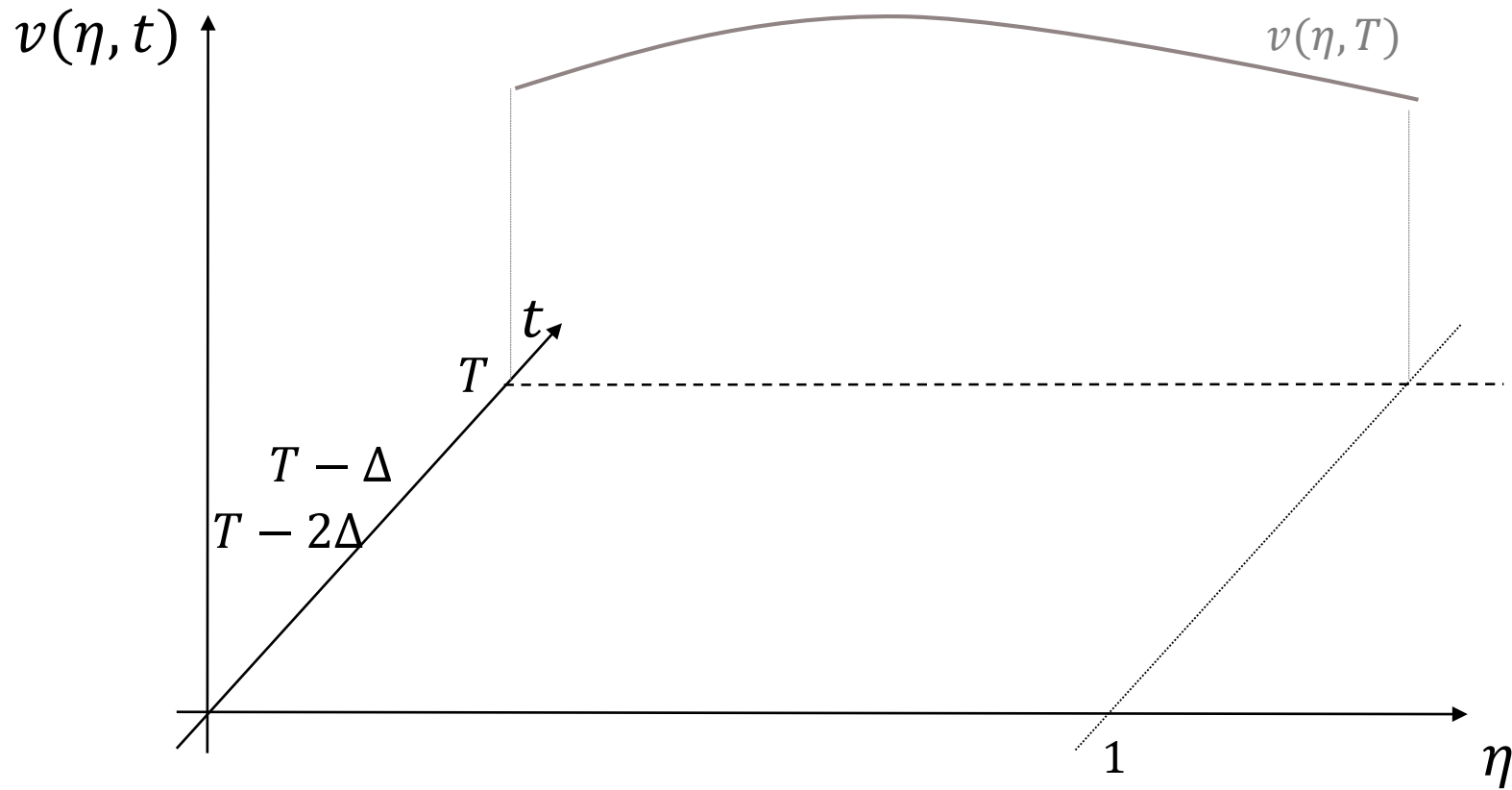
- Start at $q(0)$, solve to the right, use different procedure for two η regions depending on ψ :

- While $\psi < 1$, solve ODE for $q(\eta)$:

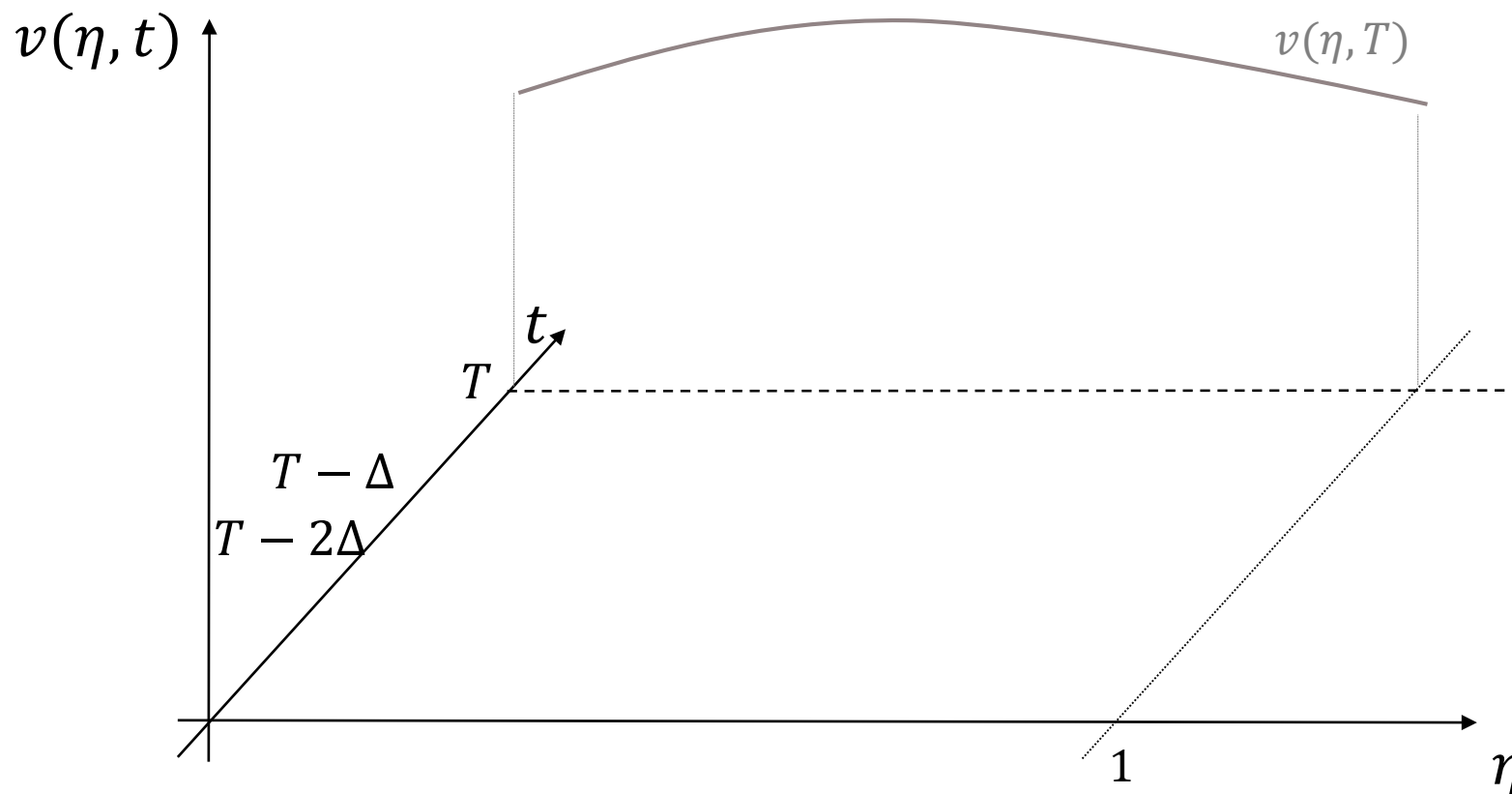
- For given $q(\eta)$, goods market clearing (4) and opt. investment (1) yield $\psi(\eta)$
- Planner conditions (2) and (3) give $(\sigma + \sigma^q)(\eta)$
- Risk equation (5) gives derivative $q'(\eta)$

- When $\psi = 1$, (2) is no longer informative, solve (1) and (4) for $q(\eta)$

4b. Value Function Iteration

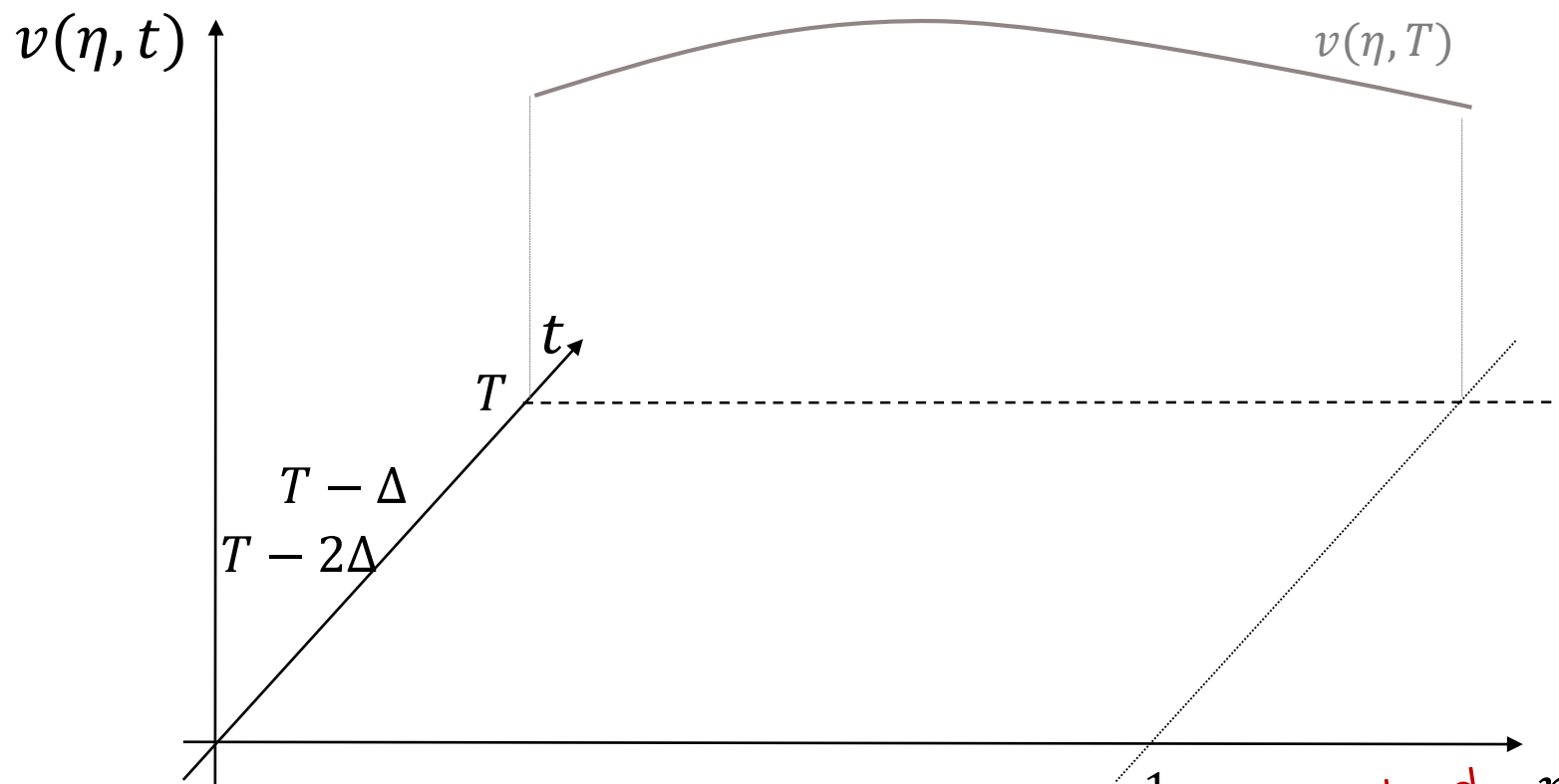


4b. Value Function Iteration



- For given $v(\eta, T)$, derive SDF ξ_T
 - Optimal investment, portfolio, consumption, at T as fcn. of η
4. Market clearing at T obtain PDE coefficient at T
 (pretend they are constant between T & $T - \Delta$)

4b. Value Function Iteration

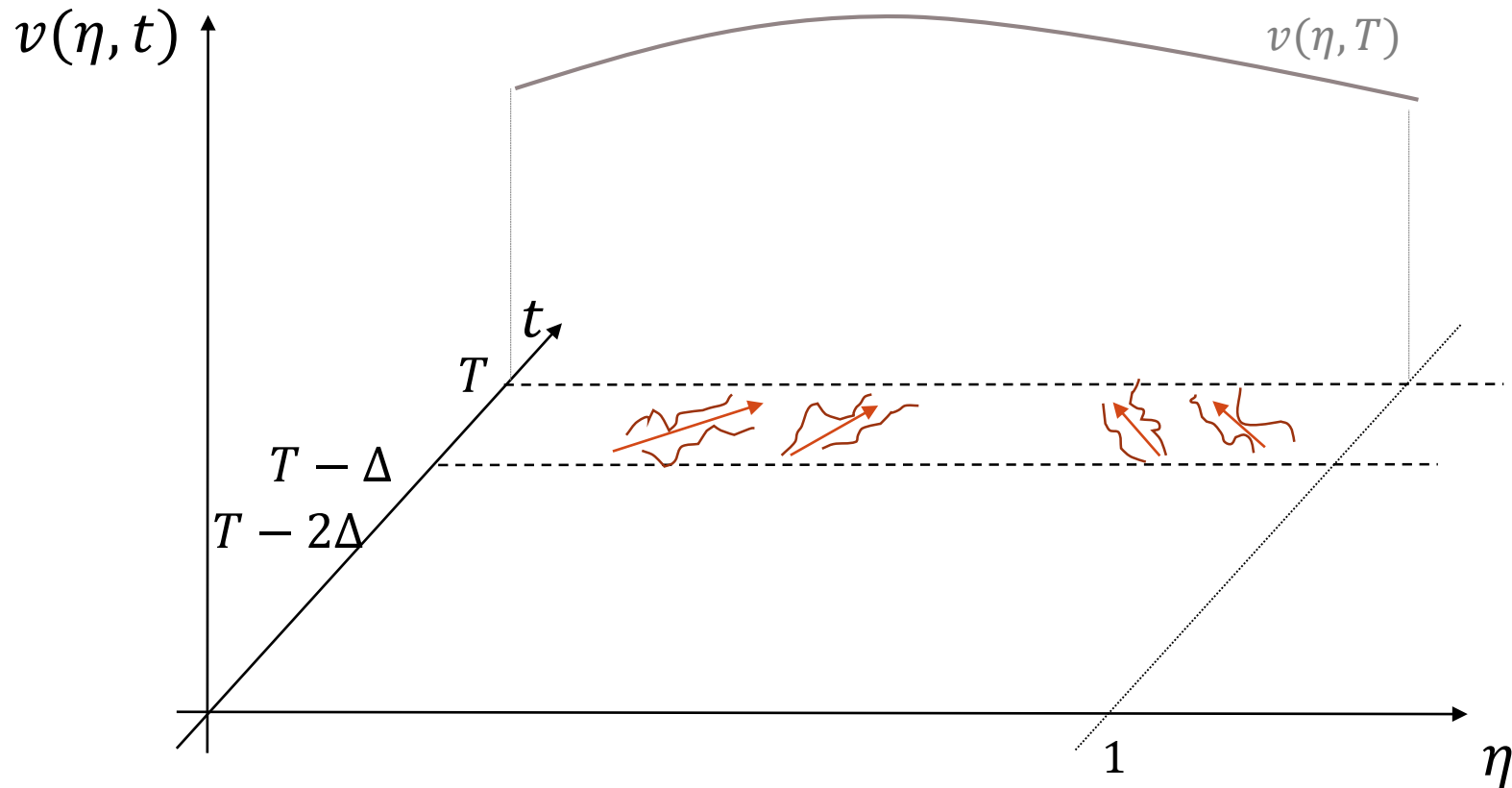


- For given $v(\eta, T)$, derive SDF ξ_T
- Optimal investment, portfolio, consumption, at T as fcn. of η

4. Market clearing at T obtain PDE coefficient at T
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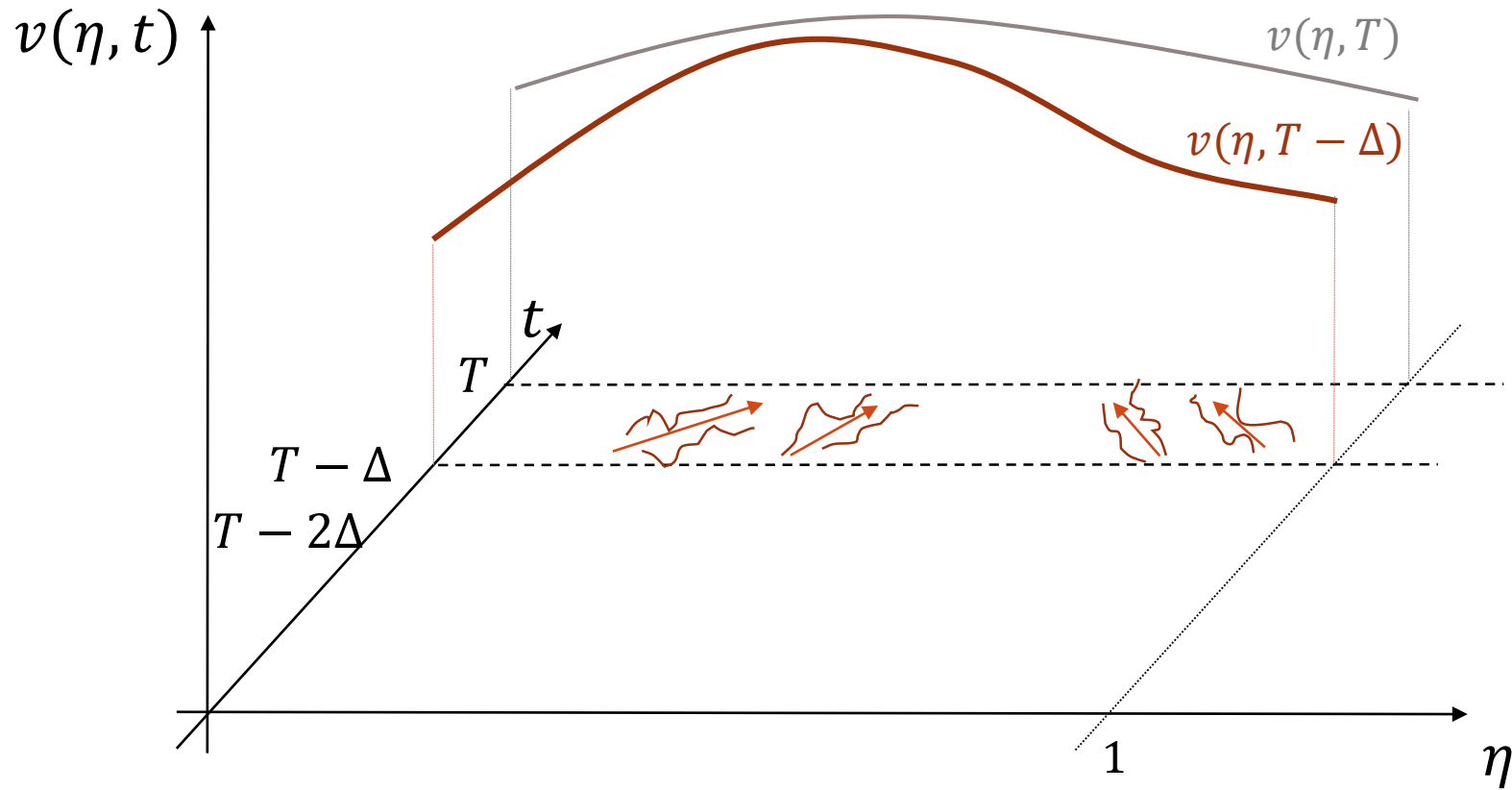
Explicit method
 Implicit method uses $T - \Delta$

4b. Value Function Iteration



- Obtain descaled value function $v(\eta, T - \Delta)$
- Repeat previous steps

4b. Value Function Iteration



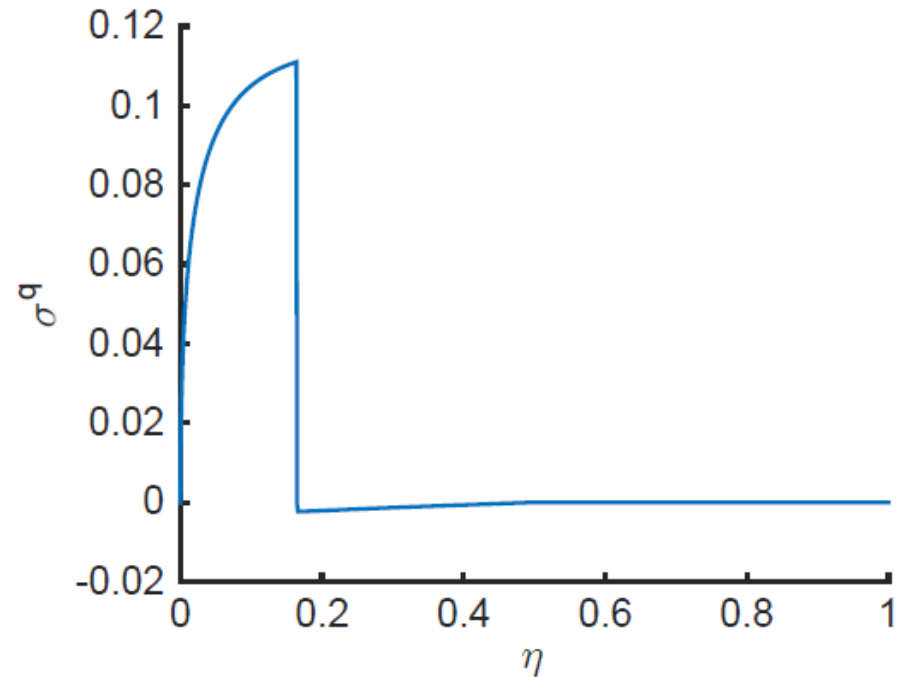
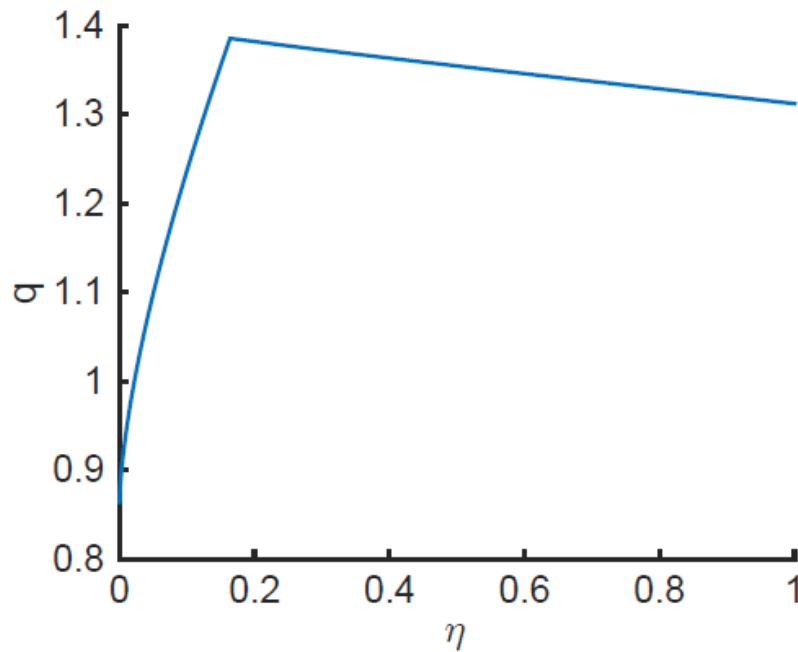
4b. Algo-code

- System of ODE (vector): solve backwards in time
- v is a vector on η -grid $[0, \Delta, 2\Delta, \dots 1]$ same for \underline{v}
- Use MatLab ODE solver to solve vector ODE
 - Sequence of steps to find $\mu^\eta, \sigma^\eta, \mu^v, \frac{c}{K}, \Phi(l) - \delta$
 - E.g. ODE45
 - Explicit vs. Implicit ODE solver
 - Explicit
 - Spatial step for ODE is Δ
 - To be stable time step $dt = o(\Delta^2)$ for explicit solver (1000 steps 10^6 time steps)
 - Implicit
- How to obtain v' and v'' from grid points
 - $$v''(n) = \frac{v(n+1) - 2v(n) + v(n-1)}{\Delta^2}$$
 - $$v'(n) = \begin{cases} \frac{v(n+1) - v(n)}{\Delta} & \text{for } \mu^\eta \eta > 0 \\ \frac{v(n) - v(n-1)}{\Delta} & \text{for } \mu^\eta \eta < 0 \end{cases}$$
- Solve $\mu_t^v v_t = v'_t \mu_t^\eta \eta_t + \frac{1}{2} v''(\sigma_t^\eta \eta_t)^2$ using Yuliy's growth CODE

||| Solution

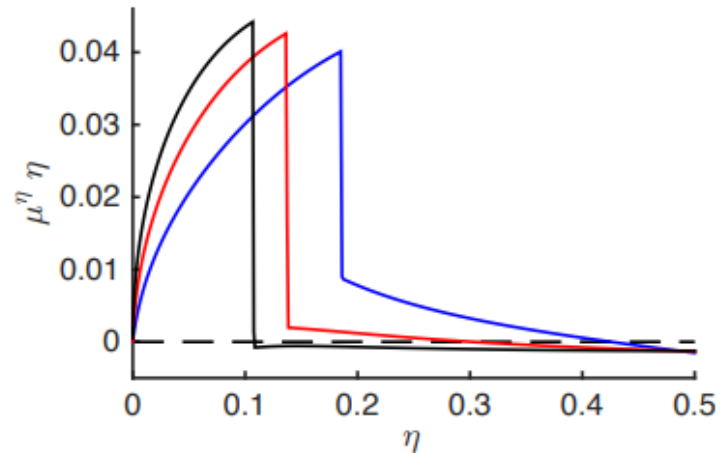
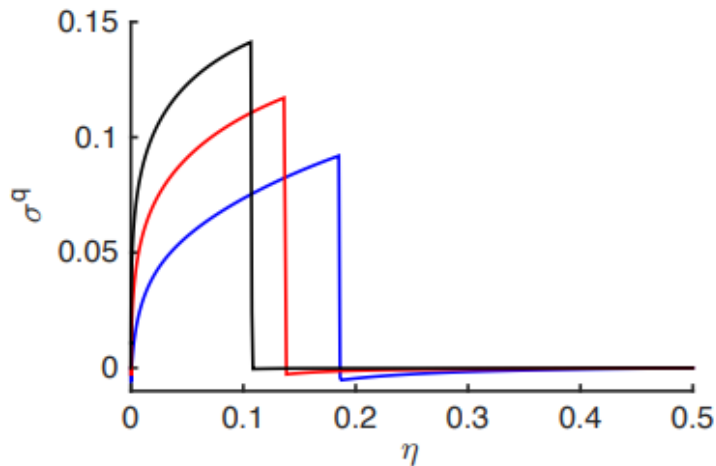
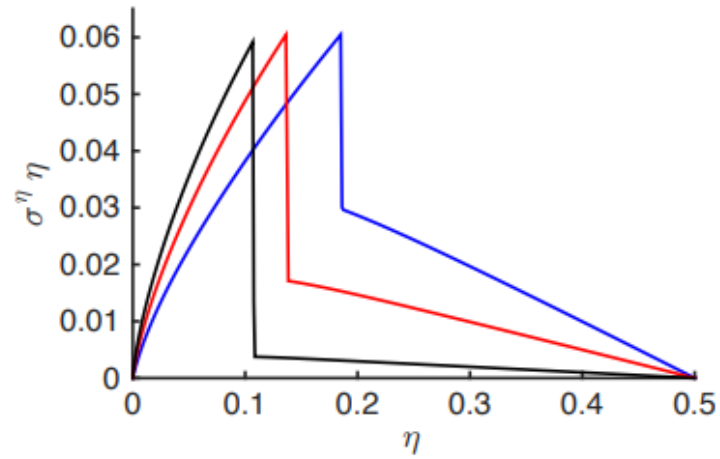
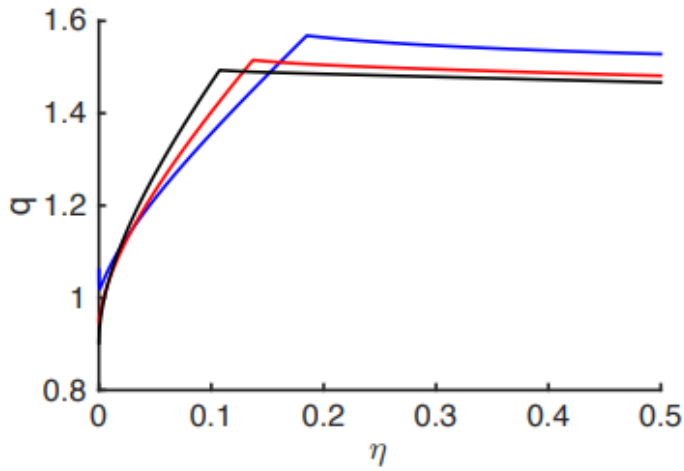
- Price of capital

Amplification



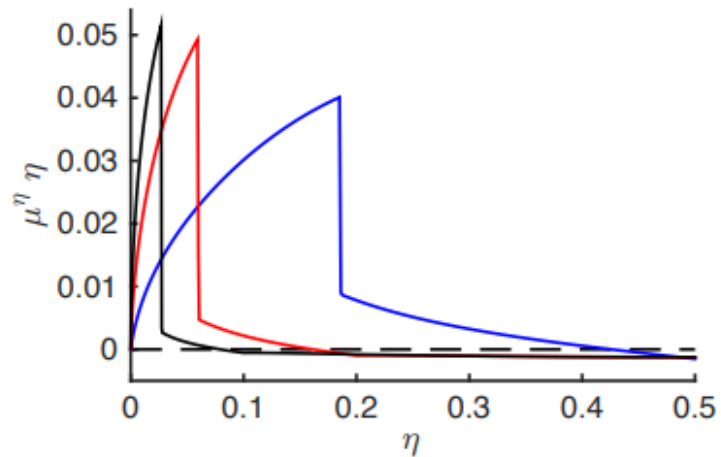
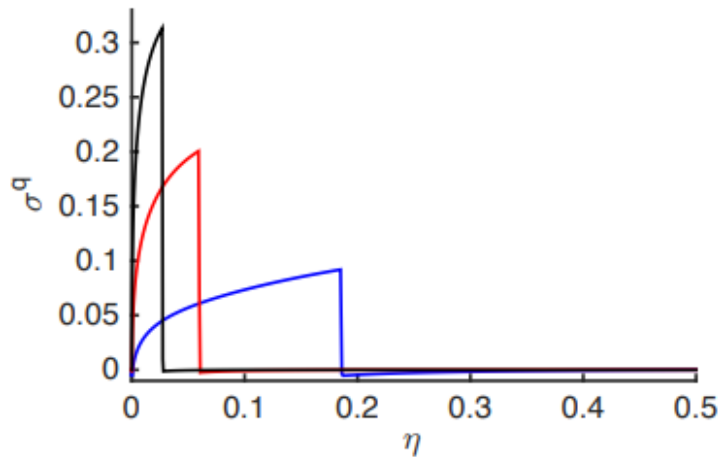
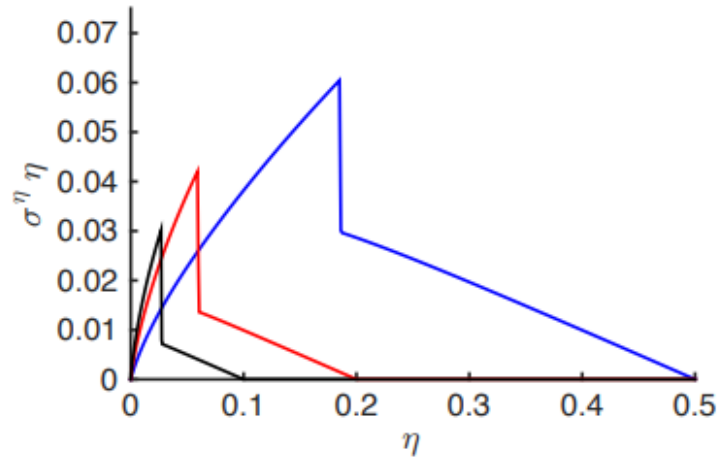
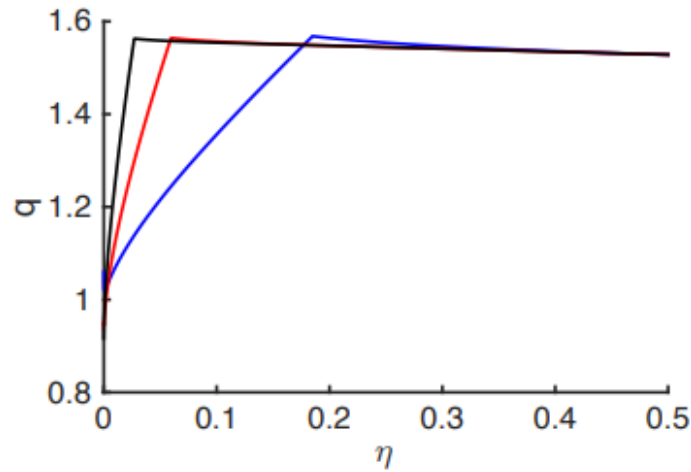
Volatility Paradox

- Comparative Static w.r.t. $\sigma = .1, .05, .01$



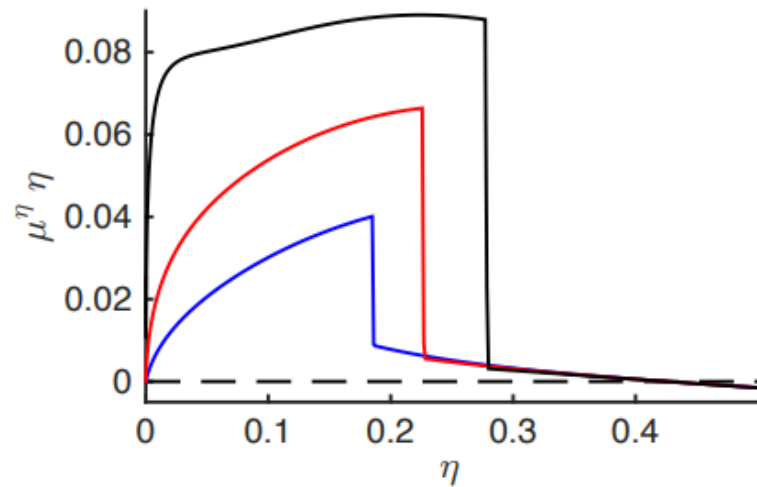
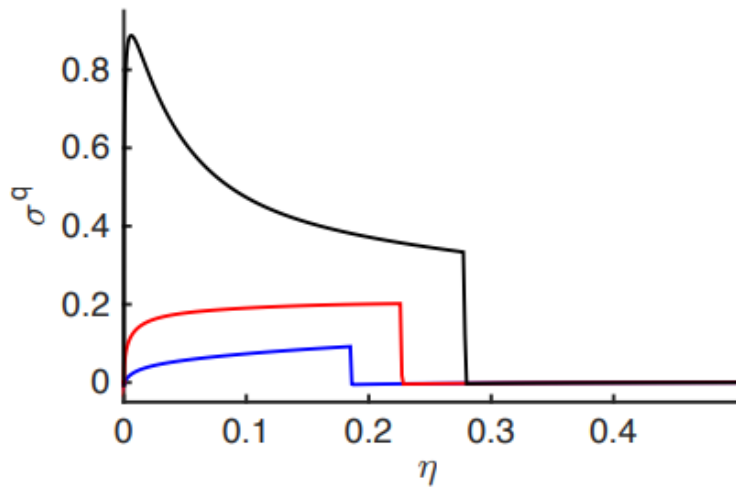
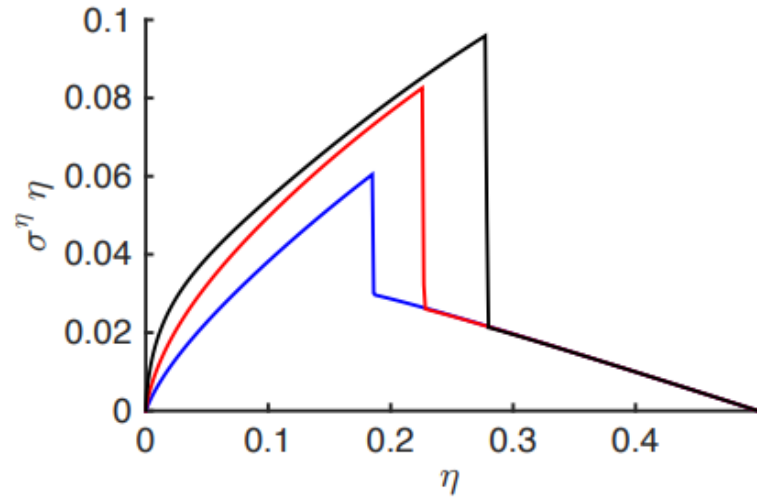
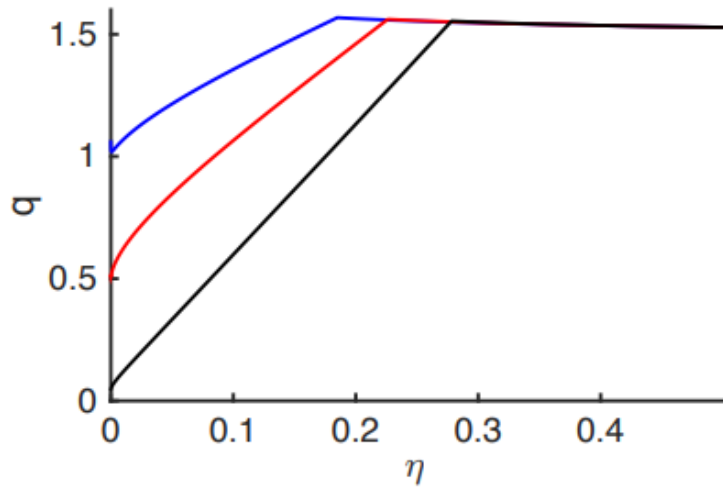
Risk Sharing via Outside Equity

- Comparative Static w.r.t. Risk sharing $\alpha = .5, .2, .1$ (skin the game constraint)



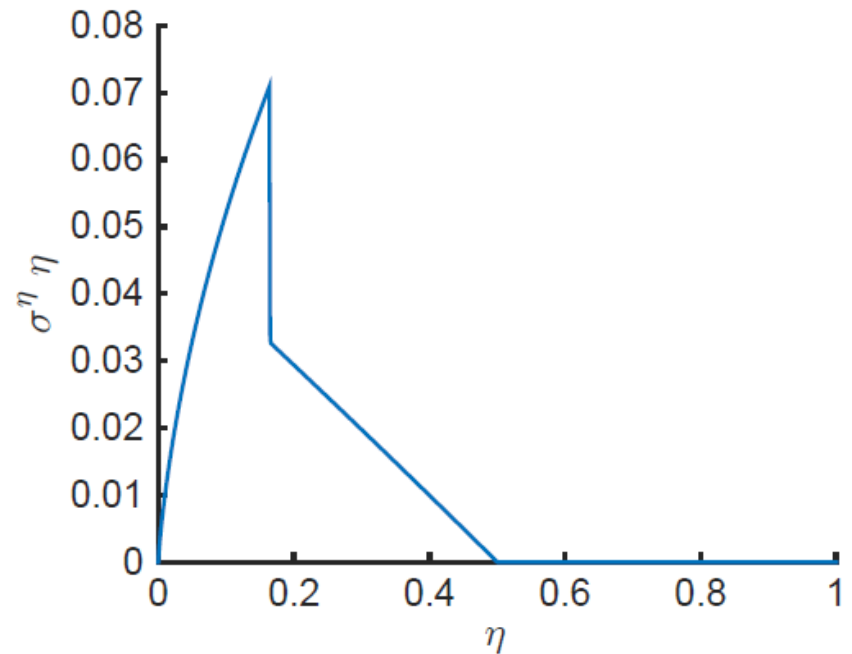
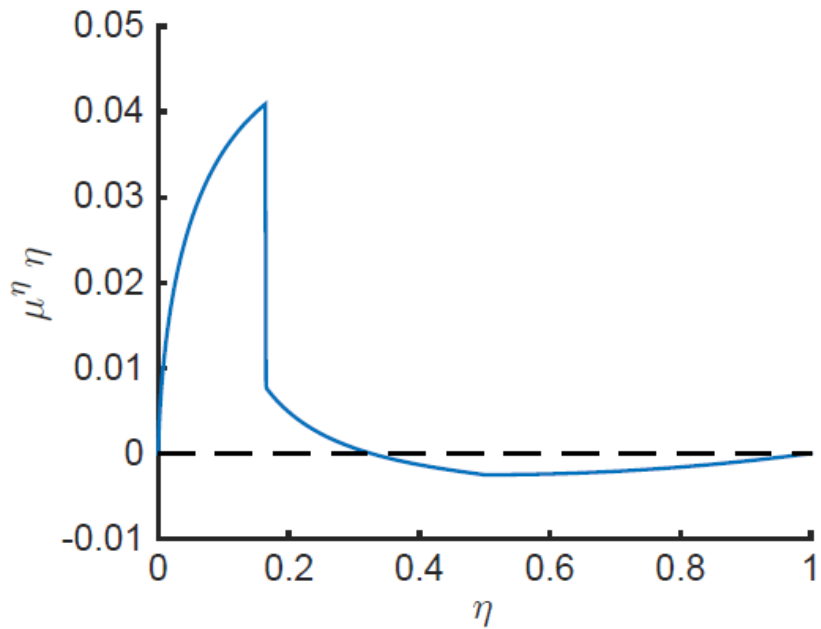
Market Liquidity

- Comparative static w.r.t. $\underline{a} = .03, -.03, -.09$



From $\mu^\eta(\eta)$ & $\sigma^\eta(\eta)$ to Stationary Distribution

- Kolmogorov forward equation



- Obtain stationary distribution

Kolmogorov Forward Equation

- Given an initial distribution $f(\eta, 0) = f_0(\eta)$, $f(\eta, t)$ satisfies the following PDE

$$\frac{\partial f(\eta, t)}{\partial t} = \frac{\partial [f(\eta, t)\mu(\eta)]}{\partial \eta} + \frac{1}{2} \frac{\partial^2 [f(\eta, t)\sigma^2(\eta)]}{\partial \eta^2}$$

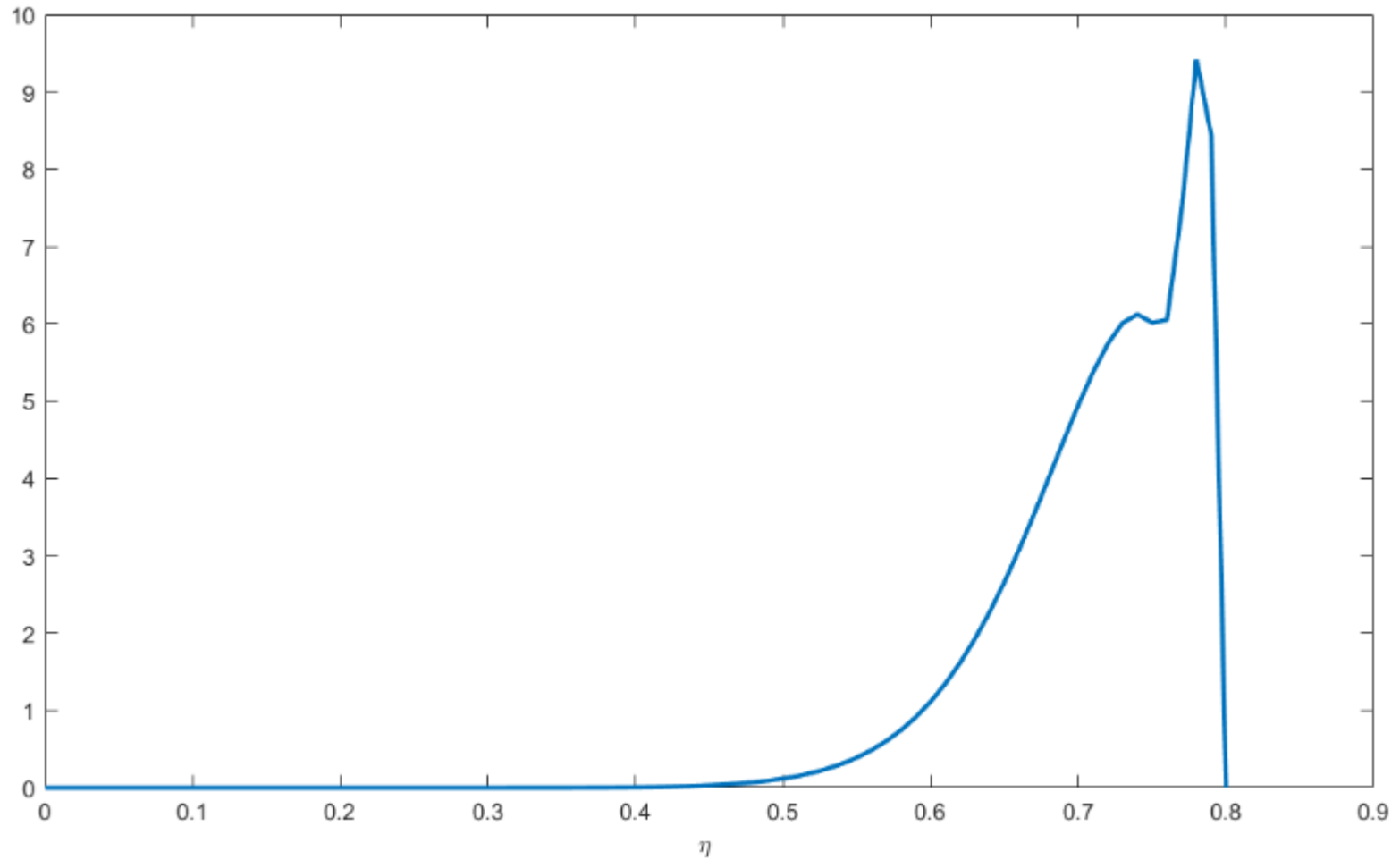
- “Kolmogorov Forward Equation” is in physics referred to as “Fokker-Planck Equation”

- Corollary: if stationary distribution $f(\eta)$ exists, it satisfies the ODE

$$0 = \frac{\partial [f(\eta, t)\mu(\eta)]}{\partial \eta} + \frac{1}{2} \frac{\partial^2 [f(\eta, t)\sigma^2(\eta)]}{\partial \eta^2}$$

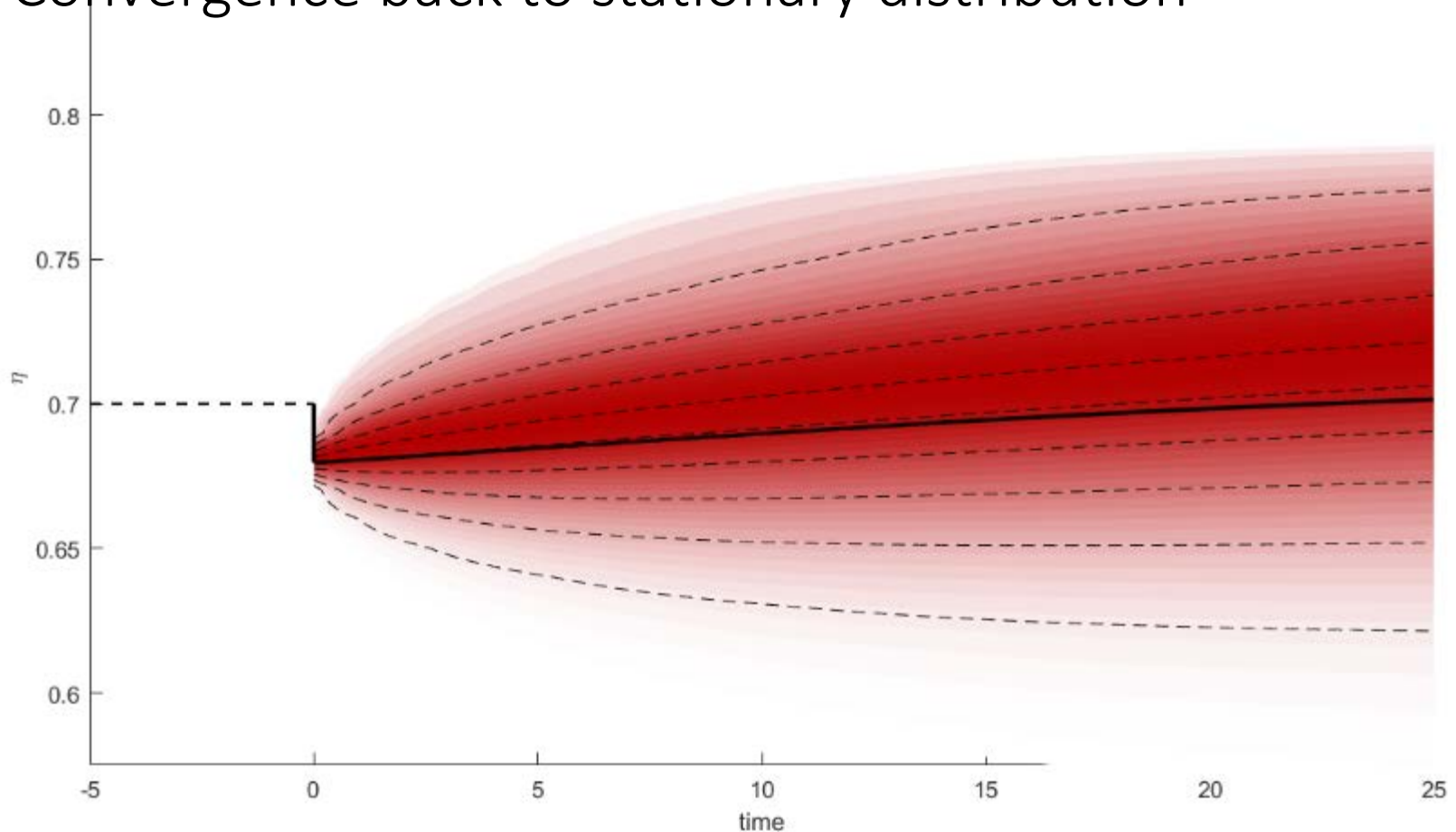
Stationary Distribution

- For different parameter settings



Fan chart

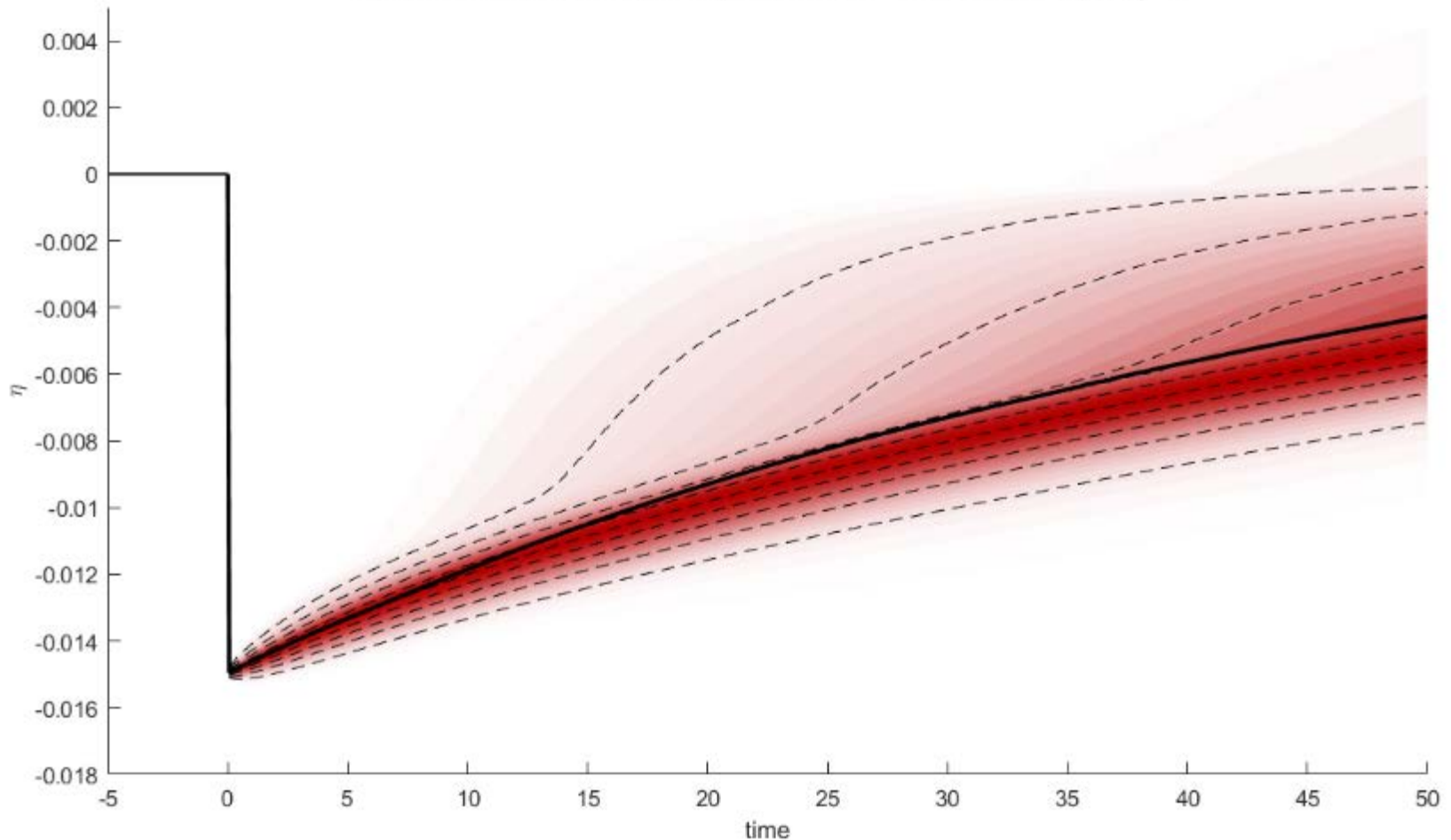
- ... the theory to Bank of England's empirical fan charts
- Start at $\eta_0 = .7$ and suffer a shock by one standard dev.
- Convergence back to stationary distribution



Distributional Impulse Response

- Difference between path with and without shock
- Difference converges to zero in the long-run

Distributional Impulse Response (Difference to Unshocked Path) at $\eta_0=0.7$



|| Solving MacroModels Step-by-Step

0. Postulate aggregates, price processes & obtain return processes
1. For given SDF processes *static*
 - a. Real investment ι , (portfolio θ , & consumption choice of *each agent*)
 - *Toolbox 1*: Martingale Approach
 - b. Asset/Risk Allocation *across types/sectors* & asset market clearing
 - *Toolbox 2*: “price-taking social planner approach” – Fisher separation theorem
2. Value functions *backward equation*
 - a. Value fcn. as fcn. of individual investment opportunities ω
 - *Special cases*
 - b. De-scaled value fcn. as function of state variables η
 - *Digression*: HJB-approach (instead of martingale approach & envelop condition)
 - c. Derive ζ price of risk, C/N -ratio from value fcn. envelop condition
3. Evolution of state variable η *forward equation*
 - *Toolbox 3*: Change in numeraire to total wealth (including SDF)
 - (“Money evaluation equation” μ^ϑ)
4. Value function iteration & goods market clearing
 - a. PDE of de-scaled value fcn.
 - b. Value function iteration by solving PDE

Extra Slides



Recent Macro-finance Literature (in cts. time)

■ Core

- BrunSan (2014), Basak & Cuoco (1998) He & Krishnamurthy (2012,13), DiTella (2013), Isohätälä et al. (2014)

■ Intermediation/shadow banking

- Phelan (2014), Adrian & Boyarchenko (2012,13), Huang (2014), Moreira & Savov (2014), Klimenko & Rochet (2015)

■ Quantification

- He & Krishnamurthy (2014), Mittnik & Semmler (2013)

■ International

- BruSan (2015), Maggiori (2013)

■ Monetary

- “The I Theory of Money” (2012), Drechsler et al. (2014)

■