

# CHILE

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**Application:** How does wealth distribution affect market quality (info efficiency, liquidity, trading volume, welfare)?

**CHILE** is uniquely suited, as one needs a model with

- Wealth effects
- Heterogeneity
- Asymmetric information



# The model. Baseline setup. “Discrete economy”

- $t \in \{1, 2\}$ .
- Risk-free asset,  $R_f = 1$ . Risky asset pays off  $\exp(v)$ ,  $v \sim N(0, \tau_v^{-1})$
- $n$  traders of size (mass)  $m = 1/n$ .
  - ▶ Trader  $a$  lives in  $[a, a + m)$
  - ▶ Observes signal  $\Delta s(a) = v \cdot m + \frac{1}{\sqrt{t(a)}} \int_a^{a+m} dB$ , precision =  $t(a) \cdot m$

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## The model. Baseline setup. “Continuum economy”

- $t \in \{1, 2\}$ .
- Risk-free asset,  $R_f = 1$ . Risky asset pays off  $\exp(v)$ ,  $v \sim N(0, \tau_v^{-1})$
- Continuum of traders  $a \in [0, 1)$ 
  - ▶ Trader  $a$  lives in  $[a, a + da)$
  - ▶ Observes signal  $ds(a) = v \cdot da + \frac{1}{\sqrt{t(a)}} dB$ , precision =  $t(a) \cdot da$

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  - ▶ Ignores his market impact by excluding himself from mkt clr. Solves

$$\max_{x(p, \Delta s(a))} E[u(W_0(a) + x(\cdot)(R - 1); a)] \quad (1a)$$

$$\text{s.t.: } \sum_{j \neq i} x_j(p, \Delta s_j) = 0. \quad (1b)$$

- **Rich heterogeneity:**  $\{W_0(a), t(a), u(\cdot, a)\}$ , arbitrary functions of  $a \in [0, 1)$ . General utilities.

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- **Rich heterogeneity:**  $\{W_0(a), t(a), u(\cdot, a)\}$ , arbitrary functions of  $a \in [0, 1)$ . General utilities.
- Log-linear equilibrium. Let  $p = \log P$ . The dollar demand of trader  $a$  is

$$dx(a) = \alpha(a)da + \beta(a)ds(a) - \gamma(a)pda.$$

# Preview of the results

## **Framework**

- Tractable, log-linear equilibrium. Closed-form solutions.
- Closed-form solutions for info efficiency, liquidity, volume, and welfare.
- Invariant relationship linking info efficiency (harder to measure) to liquidity and volume (easier to measure).
- (Money-metric) welfare can be expressed via liquidity and volume

## **Application: wealth distribution and market quality**

- Inequality is bad for info efficiency
- Inequality is good for liquidity, volume
- Ambiguous effect on welfare



# Extensions and ongoing work

## Some extensions:

- Information acquisition at  $t = 0$ . Appendix D.
- General payoffs  $V(v)$  ,  $v \sim N(0, \tau_v^{-1})$  . Appendix C.

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## Ongoing work:

- Kyle'89 setup. Traders do not ignore price impact in the discrete economy.
- Discriminatory price auction with heterogenous info
- Multi-asset model
- Dynamic, continuous-time CHILE
- ...

## Note on ignoring market impact

- Each trader assumes he has no impact on the price
- No impact on price level (Kyle's  $\lambda = 0$ ); no impact on info content of price ( $\text{cov}(ds, p) = 0$ ).

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- Each trader assumes he has no impact on the price
- No impact on price level (Kyle's  $\lambda = 0$ ); no impact on info content of price ( $\text{cov}(ds, p) = 0$ ).
- This is a small mistake. Given a trade  $dx(a)$  price changes by  $\lambda dx$ , where  $\lambda$  is finite. Similarly,  $\text{cov}(ds, p) \sim da$
- These small mistakes aggregate and do not wash away
- Equilibrium is well defined even without noise traders

## Key technical result: aggregation lemma

- Trader  $a$  observes  $\Delta s(a) = vm + 1/\sqrt{t(a)}\Delta B(a)$ . His demand is  $x(\Delta s, m)$

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$$\begin{aligned}x(ds, da) &= x_s ds + 1/2 x_{ss} ds^2 + x_m da + 1/2 x_{mm} da^2 + x_{ms} dad s \\ &= x_s ds + 1/2 x_{ss} da/t(a) + x_m da\end{aligned}$$

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- Key: non-linear term  $ds^2$  becomes non-stochastic.

$$\begin{aligned}x(ds, da) &= x_s ds + \left( \frac{x_{ss}}{2t(a)} + x_m \right) da \implies \\ \sum_{a < y} x(ds, da) &\rightarrow \int_0^y x_s ds + \int_0^y \left( \frac{x_{ss}}{2t(a)} + x_m \right) da\end{aligned}$$

**Lemma.** Suppose that demands  $x(p, \Delta s, m, a)$  are well-behaved. Then aggregate demand  $\sum_i x(\cdot)$  converges to the **Ito process**.

## Key technical result: aggregation lemma

**Lemma.** Suppose that demands  $x(p, \Delta s, m, a)$  are well-behaved. Then aggregate demand  $\sum_i x(\cdot)$  converges to an Ito process

$$dX = \beta(p, a)ds(a) + \text{drift}(p, a)da, \text{ where}$$

$$\beta(p, a) = x_s \text{ and } \text{drift}(p, a) = \frac{1}{2t(a)}x_{ss} + x_m.$$

The aggregation lemma is key to the tractability of our analysis.

- Aggregate demand is linear in  $ds(a) \Rightarrow$  the equilibrium is always generalized linear.

$$\text{Price} \propto \int_0^1 \beta(p, a)ds(a) = \nu \int_0^1 \beta(p, a)da + \int_0^1 \frac{\beta(p, a)dB(a)}{\sqrt{t(a)}}$$

- Log-normal distribution yields even more tractability:  $\beta(p, a) = \beta(a)$  and  $\text{diffusion}(p, a) = \alpha(a) - \gamma(a)p$

# Equilibrium

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Let  $\rho(a)$  be absolute risk aversion,  $\rho(a) = -u''(W_0(a), a)/u'(W_0(a), a)$ .

**Theorem.** There exists a unique equilibrium.

$dx(a) = \alpha(a)da + \beta(a)ds(a) - \gamma(a)pda$ , where

$$\beta(a) = \frac{t(a)/\tau}{\rho(a)\text{Var}[R|p]}$$

with  $\text{Var}[R|p] = \exp(\tau^{-1}) - 1$  and  $\tau = \tau_v + \tau_p$ , where  $\tau_p$  is the equilibrium price informativeness,

$$\tau_p = \frac{\left(\int_0^1 \frac{t(a)}{\rho(a)} da\right)^2}{\int_0^1 \frac{t(a)}{\rho(a)^2} da}.$$

Other coefficients are given in the closed form in the paper.

- Note: closed-form solutions, with non-CARA and rich heterogeneity!

## Wealth distribution and information efficiency: first pass

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**Proposition.** Info efficiency is given by

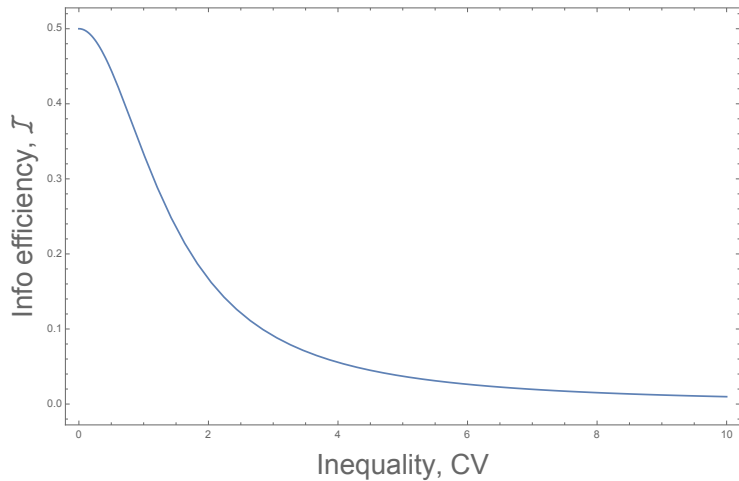
$$\mathcal{I} = \frac{\bar{t}}{\bar{t} + \tau_v(1 + CV^2)}, \text{ where}$$

$CV$  = standard deviation of wealth/average wealth

is a *coefficient of variation*.

There is a negative relationship between inequality ( $CV$ ) and information efficiency ( $\mathcal{I}$ ).

# Inequality and info efficiency.



# Inequality and info efficiency. Intuition

- Price reflects the weighted average of signals.  $p \propto \int \beta(a) ds(a)$
- Weights  $\propto \beta(W_0) \propto W_0$
- More weight on wealthier traders
- What is more informative:  $0.5(v + \epsilon_1) + 0.5(v + \epsilon_2)$  or  $0.99(v + \epsilon_1) + 0.01(v + \epsilon_2)$ ?

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- What is more informative:  $0.5(v + \epsilon_1) + 0.5(v + \epsilon_2)$  or  $0.99(v + \epsilon_1) + 0.01(v + \epsilon_2)$ ?
- Key effect: whose signal noise is reflected more in prices?  
$$p \propto \int \beta(a) ds(a) = v \int \beta(a) da + \int \frac{\beta(a)}{\sqrt{t(a)}} dB(a)$$

Absent in LE a-la Hellwig (1980), signal noise is washed out by LLN

## Wealth distribution and information efficiency

Suppose that  $t(a) = \bar{t}$ , and all traders are CRRA with the same RRA.

**Corollary.**

$$\mathcal{I} = \frac{\bar{t}}{\bar{t} + \tau_v(1 + CV^2)} \leq \frac{\bar{t}}{\bar{t} + \tau_v},$$

maximum  $\mathcal{I}$  is attained when  $CV = 0$ .

Maximum info efficiency is attained when there is no inequality.

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**Intuition.**

- Price cannot reflect more than one could infer by seeing each signal:  
 $\tau_p \leq \int_0^1 t(a) da$
- Suppose we have  $s_1 = v + \frac{1}{\sqrt{t_1}}\epsilon_1$  and  $s_2 = v + \frac{1}{\sqrt{t_2}}\epsilon_2$  ( $\epsilon_i$  are standard normal). Known result:  $\{s_1, s_2\}$  is info equivalent to  $s = t_1 s_1 + t_2 s_2$
- The best way to aggregate signals is with weights proportional to precisions.

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- The best way to aggregate signals is with weights proportional to precisions. If  $\beta(a) \propto t(a)$ ,  $\tau_p = \int_0^1 t(a) da$
- But we have weights  $\propto \beta(a) \propto t(a)/\rho(a)$ .
- $\beta(a) \propto t(a)$  iff  $\rho(a) = \bar{\rho}$  which is only possible when  $W_0(a) = \bar{W}_0$



# Wealth distribution and information efficiency: general case

Do our results still hold when utilities are heterogenous and non-CRRA?

When precisions are heterogenous?

When iprecisions are endogenous?

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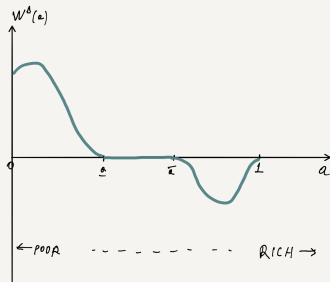
**Definition.** Gateaux derivative  $\mathcal{I}'(W_0(a))[W_0^\Delta(a)]$  in the direction  $W_0^\Delta(a)$  is

$$\mathcal{I}'(W_0(a))[W_0^\Delta(a)] = \lim_{\epsilon \rightarrow 0} \frac{\mathcal{I}(W_0(a) + \epsilon W_0^\Delta(a)) - \mathcal{I}(W_0(a))}{\epsilon}$$

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Assume  $W_0(a) \uparrow$  in  $a$  (WLOG)

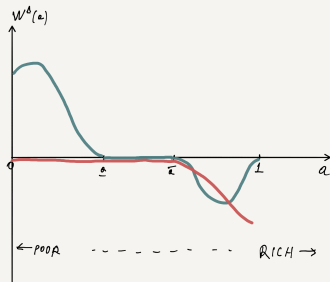
**Definition.** Robin Hood variation is a direction  $W_0^\Delta(a) \neq 0$  such that  $W_0^\Delta(a) \geq 0$  for  $a < \underline{a}$  and  $W_0^\Delta(a) \leq 0$  for  $a > \bar{a}$ .



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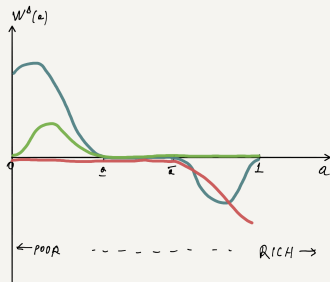
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# Wealth distribution and information efficiency: general case

How  $\mathcal{I}$  changes when  $W_0(a)$  changes? When  $t(a)$  changes?

## The sequence of exercises:

1. Vary  $W_0(a)$  keeping  $t(a)$  fixed
2. Vary  $t(a)$  keeping  $W_0(a)$  fixed
3. Vary both

## Wealth distribution and information efficiency: general case

**Proposition.** Assume DARA utility, exogenous precisions+technical conditions. There exists  $0 < a^* < 1$  such that for all Robin Hood  $W^\Delta(a)$  with  $\underline{a} \leq a^* \leq \bar{a}$ :

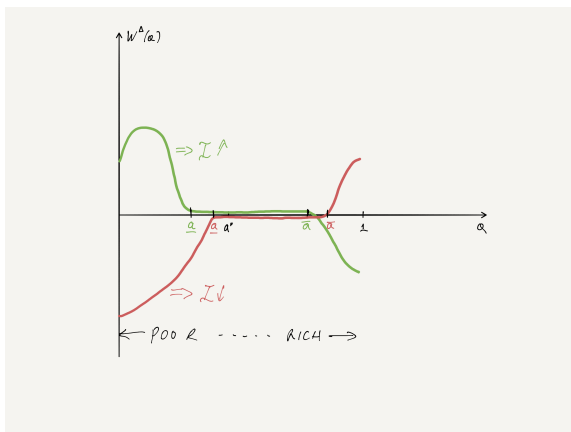
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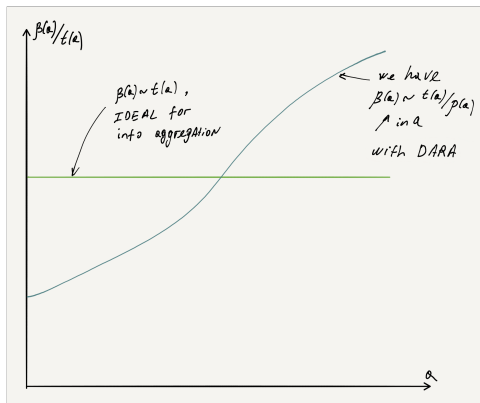
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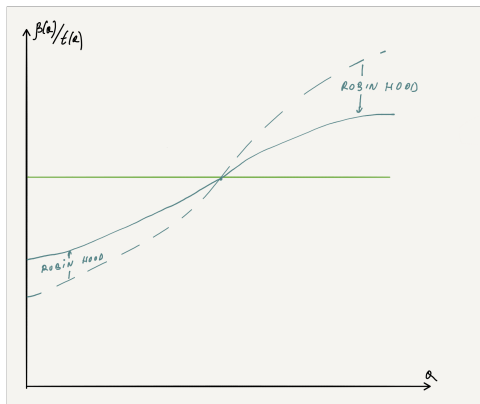
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# Wealth distribution and information efficiency

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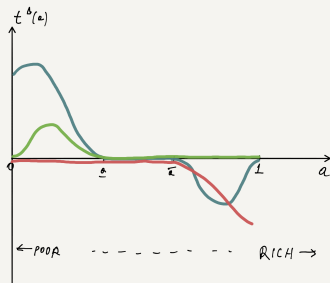
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# Distribution of precisions and information efficiency

Assume  $W_0(a) \uparrow$  in  $a$  (WLOG). Here, we will vary  $t(a)$  keeping  $W_0(a)$  fixed.

**Definition.** Robin Hood variation of  $t(a)$  is a direction  $t^\Delta(a) \neq 0$  such that  $t^\Delta(a) \geq 0$  for  $a < \bar{a}$  and  $t^\Delta(a) \leq 0$  for  $a > \bar{a}$ .



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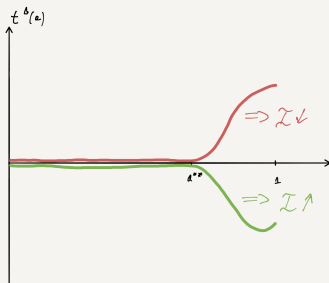
## Intuition

- Rich trade too aggressively, poor too passively
- Giving less info to the rich makes them less aggressive; analogously, for poor

# Distribution of precisions and information efficiency

**Corollary.** Assume DARA utility+technical conditions. For any  $t^\Delta(a) \neq 0$  such that  $t^\Delta(a) \geq 0$  for  $a > a^{**}$  and  $t^\Delta(a) = 0$  otherwise:

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## Implication

- MIFID (unbundling of research fees and trading commissions) makes it harder for small funds to acquire info compared to large. Potentially detrimental effects for info efficiency

# Distribution of wealth and information efficiency with endogenous information

**Proposition.** Assume DARA utility, info cost( $t$ ) =  $t^c$ ,  $c > 1$ , technical conditions. There exists a unique overall equilibrium. There exists  $0 < a^* < a^{**} < 1$  such that for all Robin Hood  $W^\Delta(a)$  with  $\underline{a} \leq a^* < a^{**} \leq \bar{a}$ :

$$\mathcal{I}'[W_0^\Delta(a)] > 0, \quad \mathcal{I}'[-W_0^\Delta(a)] < 0.$$

- Combination of two previous exercises
- Robin Hood variation:
  - ▶ Flattens the distribution of risk tolerances
  - ▶ Decreases the info of the rich and increases the info of the poor via endogenous info acquisition (rich acquire more info)

# Information efficiency, liquidity and volume

## Definition

- Liquidity  $\mathcal{L} = \int_0^1 \gamma(a) da$
- Volume  $\mathcal{V}^2 = \int_0^1 dx(a)^2$

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## Key underlying equation

- To have small equilibrium demands, must have demand elasticity  $\sim da$
- Suppose price  $\downarrow$  by 1%. Asset is cheaper, demand  $\uparrow$  (cost component). Perhaps fundamental is lower, demand  $\downarrow$  (info component).

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**Cost component** = information component +  $O(da)$

$$1 = \frac{\tau_p}{\tau_p + \tau_v} \cdot \frac{\int_0^1 \gamma(a) da}{\int_0^1 \beta(a) da}$$

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$$\underbrace{\mathcal{I}(1 - \mathcal{I})}_{\text{hard to measure}} = \underbrace{\frac{\mathcal{V}^2}{\sigma_v^2 \mathcal{L}^2}}_{\text{easier to measure}}.$$

## Liquidity and wealth distribution

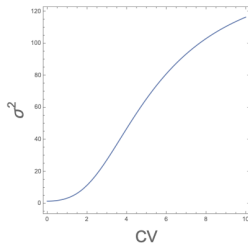
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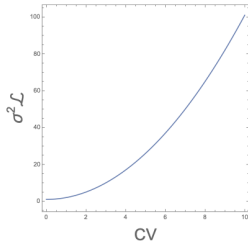
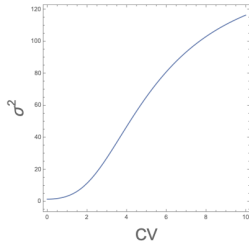
**Proposition.** The volatility  $\sigma^2 = \text{Var}[R|p]$  is increasing in  $CV$ .



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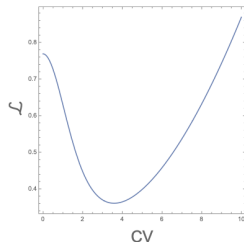
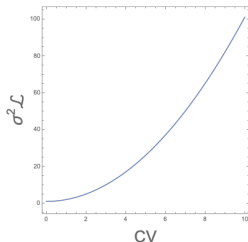
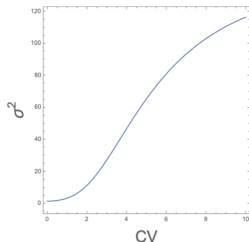
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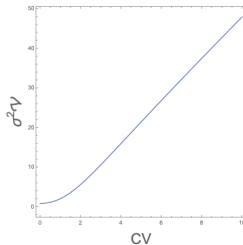
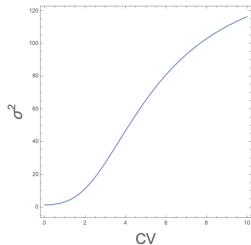
**Proposition.** The volatility  $\sigma^2 = \text{Var}[R|p]$  is increasing in  $CV$ . Risk-scaled liquidity  $\mathcal{L}\sigma^2$  is increasing in  $CV$ . Liquidity is U-shaped in  $CV$ .



# Volume and wealth distribution

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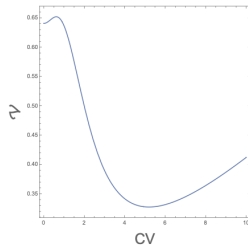
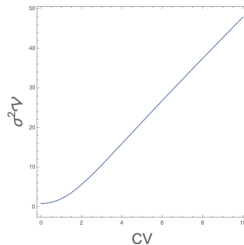
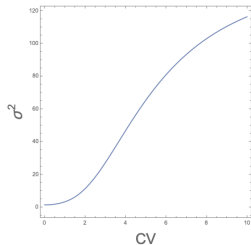
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**Proposition.** The volatility  $\sigma^2 = \text{Var}[R|p]$  is increasing in  $CV$ .  
Risk-scaled volume  $\mathcal{V}\sigma^2$  is increasing in  $CV$ . Volume is non-monotone in  $CV$ .



# Welfare and wealth distribution

## Welfare:

- Money-metric welfare loss
- How much money do we need to give to  $a$  to make him as happy as in the first best?
- Aggregate. Take the  $m \rightarrow 0$  limit

**Proposition.**  $\mathcal{W}_{loss} = \frac{\nu^2}{2\mathcal{L}}.$

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**Proposition.**  $\text{sign}(\mathcal{W}_{loss}(CV)') = \text{sign}\left(\frac{1+CV^2}{\bar{t}+\tau_v(1+CV^2)} - 1.6\right)$ .

- Using invariant relationships, can write  $\mathcal{W} = \frac{\sigma_v^2}{2}\mathcal{I}(1-\mathcal{I})\mathcal{L}$
- Large info frictions (small  $\bar{t}$ , or small  $\tau_v$ ): inequality is bad for welfare. (Effect of inequality on  $\mathcal{I}$  dominates)
- Small info frictions (large  $\bar{t}$ , or large  $\tau_v$ ): inequality is good for welfare. (Effect of inequality on  $\mathcal{L}$  dominates)

# Conclusion

- A new heterogeneous information asset pricing framework
- Tractable. General utilities. Rich heterogeneity. Closed-form solutions
- Allows to analyse how wealth distribution affects market quality
- Active follow-ups:
  - ▶ Kyle in CHILE
  - ▶ Discriminatory price auction/ static limit order book
  - ▶ Continuous-time CHILE
  - ▶ Multi-asset CHILE
  - ▶ ...



# How we solved for equilibrium

1. Consider a discrete economy where each trader  $a$  believes other traders' demands are  $d\hat{x}(b) = \hat{\alpha}(b, m) + \hat{\beta}(b, m)ds(b) - \hat{\gamma}(b, m)p$

Solves

$$x^{BR}(a, \Delta s, p, m) = \arg \max_{x(p, \Delta s(a))} E[u(W_0(a) + x(\cdot)(R - 1); a)] \quad (2a)$$

$$\text{s.t.: } \int_{-a} d\hat{x}(b) = 0. \quad (2b)$$

2. Use aggregation lemma to compute  $\lim_{m \rightarrow 0} \sum x^{BR}(a, p, m)$ . (Involves implicitly differentiating FOC to get  $x_s$ ,  $x_{ss}$  etc)
3. Require consistency
  - ▶  $\lim_{m \rightarrow 0} d\hat{x} = \alpha(b) + \beta(b)ds(b) - \gamma(b)p$
  - ▶  $\lim_{m \rightarrow 0} dx^{BR} = \alpha(b) + \beta(b)ds(b) - \gamma(b)p$
4. We've shown this procedure yields the limiting equilibrium in the discrete economy

# Technical conditions

Technical conditions = x-sectional distribution of wealth (relative risk aversion) has unbounded (compact) support. [▶ back](#)