The Analytics of Monetary Shocks with Generalized Hazard Functions

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- Caballero-Engel introduced Generalized Hazard Function (GHF)
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 - Max CIR (kurtosis): Constant GHF (Calvo), Min: Golosov-Lucas.
 - ► Flexibility Index ≠ summary of IRF.

Firm's Problem

► Price gap: *x* deviation from ideal markup (no inflation) follows driftless random walk: $dx = \sigma dW$ if price not changed

Second order approximation to profit function B x², Bx² cost of deviation x.

► Firm can always pay fixed cost Ψ and adjust prices

- Each period w/prob. κ dt firm draws cost ψ ∼ G(·) assume (WLOG) that G : [0, Ψ] → [0, 1].
- Firm minimized expected discounted (at rate r) cost
- Optimal decision rule: sS bands $\bar{x}(\psi)$ increasing in ψ , and $\pm X$. (Caballero and Engel)
- Symmetry: optimal return pint to x = 0, and bands <u>x(ψ)</u> = −x̄(ψ) and ±X

Firm's Problem: Bellman equation

 \blacktriangleright v(x) value function of firm w/price gap (markup deviation) x:

$$rv(x) = \min \{ Bx^2 + \frac{\sigma^2}{2}v''(x) + \kappa \int_0^{\Psi} \min \{ \psi + v(0) - v(x), 0 \} dG(\psi),$$

$$r(v(0) + \Psi) \}$$

- $\mathbf{v}(0)$ is the minimum, optimal markup adjustment $\mathbf{x} = \mathbf{0}$
- Random cost ψ drawn w/prob.: $\kappa dG(\psi)$ each period of length dt
- Firm adjust if $\psi \leq v(x) v(0) =$ reduction in cost
- Firm can always pay (largest) cost \u03c8 and adjust.

Firm's Problem: Bellman equation (cont.)

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$$r(v(0) + \Psi) \}$$

• Optimal decision rule X and $\bar{x}(\cdot)$:

• If |x| reaches $X \rightarrow$ adjust with certainty (mass point).

At each $|x| \rightarrow adjust w/ prob \Lambda(x) = \kappa G(v(x) - v(0))$ per dt

• Generalized Hazard Function $\Lambda(x)$:

- adjustment probability increases with |x| (Caballero-Engel),
- > at higher deviation |x| more cost are worthwhile to be paid.

• We allow G to have mass points, and also $\Psi = \infty \implies X = \infty$.

• Calvo⁺: G has a mass point at $\psi = 0$, otherwise it is constant.

Inversion Result

- Caballero Engel: CDF $G \implies \Lambda$ increasing and symmetric GHF.
- New result: Λ is lincreasing and symmetric \implies G GHF:

Fix volatility σ^2 , curvature *B*, and discount rate *r*.

Consider an upper bd X & a symmetric, weakly increasing GHF Λ . Then there is a unique Ψ and CDF G that rationalize X, Λ .

- Importance: can use any weakly increasing symmetric GHF to fit data or write models.
- Solving Bellman equation and policy is hard, non-linear o.d.e. Solving for *G* given A easy and constructive: linear o.d.e. (discrete *G*, linear eqns).

Sketch of Proof (*skip for time*). Recall for 0 < x < X:</p>

$$rv(x) = Bx^{2} + \frac{\sigma^{2}}{2}v''(x) + \kappa \int_{0}^{\Psi} \min \{\psi + v(0) - v(x), 0\} dG(\psi)$$

= $Bx^{2} + \frac{\sigma^{2}}{2}v''(x) + \kappa \int_{0}^{v(x) - v(0)} \psi dG(\psi)$
+ $\kappa [v(0) - v(x)] G(v(x) - v(0))$

▶ Differentiate value fcn: $u(x) \equiv v'(x)$, use $\Lambda(x) \equiv \kappa G(v(0) - v(x))$

$$[r + \Lambda(x)] u(x) = 2Bx + \frac{\sigma^2}{2}u''(x)$$
 for $x \in [0, X]$

- Use symmetry at x = 0 and smooth pasting at x = X to motivate boundaries for u in domain [0, X]
- Solve for *u* given Λ . Linear ode (use Sturm-Liouville, Direchlet) Show $u \ge 0$ so u = v' > 0: $\iff \Lambda'(x) = \kappa G'(v(x) - v(0))u(x)$.

• Then
$$v(x) = u'(0)\frac{\sigma^2}{2r} + \int_0^x u(z)dz$$
 for $x \in [0, X]$.

Optimal Adjustment Intensity Model

- ► Alternative model that also produces generalized hazard rate∧
- Firms chose the intensity of adjustment ℓ at a cost $c(\ell)$.
- ► Also can pay $\Psi \leq \infty$ if they want to adjust w/certainty

$$rv(x) = \min \left\{ Bx^2 + \frac{\sigma^2}{2}v''(x) + \min_{\ell \ge 0} \ell (v(0) - v(x)) + c(\ell) \right\}$$

, $r(\Psi + v(0))$

- Similar to Rational Inattention: Woodford (2007) and others.
- Solution: GHF $\Lambda(x) = \ell^*(x)$, symmetric, increasing.
- Also: any increasing $\Lambda(\cdot)$ is rationalized by a cost $c(\cdot)$.
- Technical No restriction on tail of Λ.

Steady State Statistics: Invariant Dist. & Frequency

lntermediate step: f(x) invariant distribution of price gaps

Solves the KFE for all $x \in [-X, X]$, $x \neq 0$:

 $f(x)\Lambda(x) = \frac{\sigma^2}{2}f''(x)$ and f continuous at all xwith f(-X) = f(X) = 0 and $\int_{-X}^{X} f(x)dx = 1$.

▶ *N*_a Number of price changes per unit of time:

$$N_{a} = \underbrace{2}_{symmetry} \left[\underbrace{\int_{0}^{X} f(x) \Lambda(x) dx}_{\#w/x \times \Pr \Delta p} \underbrace{-\frac{\sigma^{2}}{2} f'(X)}_{\Pr \Delta pat \pm X} \right]$$

uses symmetry of *f* around x = 0.

• Interpretation of $\Lambda(x)f(x)$ and $\frac{\sigma^2}{2}|f'(X)|$.

Figure: Density function for the Invariant distribution of gaps: f(x)





Steady State Statistics: Dist. of price changes Δp

- ▶ Price changes have mass point(s) at $\pm X$ if $X < \infty$
- Symmetric *density* of price changes q for all $x \in (-X, X)$

$$\Delta p = \begin{cases} -x & \text{w/ density } q(x) = \frac{\Lambda(x) f(x)}{N_a} \text{ provided that } |x| < X, \\ -X & \text{w/ probability mass } \frac{\frac{\sigma^2}{2} |f'(X)|}{N_a} \end{cases}$$

• CDF $Q: [-X, X] \rightarrow [0, 1]$ so that Q' = q in the interior.

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 \blacktriangleright *k*-moment of price changes (e.g. variance, kurtosis):

$$\mathbb{E}(\Delta p^{k}) = \frac{2\left[\int_{0}^{X} x^{k} \Lambda(x) f(x) dx - X^{k} \frac{\sigma^{2}}{2} f'(X)\right]}{N_{a}}$$

Examples: Power Generalized Hazard Function Λ

• A common case used it the literature is $\Lambda(x) \propto x^2$.

- Solution for f are Bessel functions.
- We consider a generalization $\Lambda(x) = \kappa \left(\frac{x}{\chi}\right)^{\nu}$ for $\nu > 0$.
- ▶ Note $\nu = 0$ is Calvo and $\nu \to \infty$ Golosov-Lucas.
- Kurt(Δp) depends on shape, controlled by ν and $\frac{\sigma^2}{2\kappa}/X^2$.
- *Kurt*(Δ*p*) of quadratic case (ν = 2) is low, about 1.75.
 More so if *X* is finite.

Example of density price changes: $q(\Delta p)$ Quadratic Hazard $\Lambda(x) = \kappa x^2$, $X = \infty$, different $\eta \equiv \left(\frac{2\kappa}{\sigma^2}\right)^{\frac{1}{4}}$ 1.00 0.75 $q(\Delta p)$ 0.50 0.25 0.00 -2 -1



Hazard function $\Lambda(x) = \kappa x^{\nu}$ and $X = \infty$

Figure: Kurtosis behavior with a power hazard function as *(skip)* s: fraction of price changes away from boundary X varies



Hazard function $\Lambda(x) = \kappa \left(\frac{x}{X}\right)^{\nu}$ and $X < \infty$

Recovering f, Λ and G (or c) from Δp dist. q

- ▶ Price changes have density $\Delta p \sim q(\cdot)$ for $x \in (-X, X)$.
- Assume q is symmetric., let Q be its CDF of q, so q = Q'.
- Using equations above: $q \implies f \implies \Lambda$:

Invariant distribution *f* fo price gaps:

$$f(x) = \frac{2}{Var(\Delta p)} \left[\int_{x}^{\infty} (1 - Q(z)) \, dz \right] \text{ for all } x \ge 0$$

and Generalized Hazard Function A:

$$\Lambda(x) = \frac{N_a \operatorname{Var}(\Delta p)}{2} \frac{q(x)}{\int_x^\infty (1 - Q(z)) \, dz} \text{ for all } x > 0$$

Given A & previous recovery results, we get:

 $q \implies f \implies \Lambda \implies G$ (distribution of menu cost) or

 $q \implies f \implies \Lambda \implies c$ (cost function)

Recovering f, Λ and G from q: trivial proof (*skip*)

Model gives us:

$$\frac{\Lambda(x)f(x)}{N_a} = q(-x) \text{ all } x \ge 0$$
$$\frac{\sigma^2}{2}f''(x) = \Lambda(x)f(x) \text{ all } x \ge 0$$
$$\sigma^2 = Var(\Delta p) N_a$$

Replace KFE into eqn for q:

$$\frac{\sigma^2}{2}\frac{f''(x)}{N_a} = q(-x) \text{ all } x \ge 0$$

- lntegrate w.r.t x twice, used expression for σ^2 gives f(x).
- Use f and definition of q again to get Λ .

Estimating distribution Q

- Cavallo's scraped data $\{\Delta p_{it}\}$: no time agg. + low meas. error.
- Mixture of distribution w/same Kurtosis has higher Kurtosis
- Unobserved heterogeneity b_i across products i within category

 $\Delta p_{it} = b_i \Delta \tilde{p}_t \text{ for } i \in I \text{ and } t \in T(i)$ $\Delta \tilde{p}_t \sim Q \text{ and } b_i \perp \Delta \tilde{p}_t$

- Theory: price changes iid across t for a given product.
- Non-parametrically identify common distribution *Q* for a category variation on Kotlarski's lemma (non-parametric random effects).
- Estimate Kurtosis, $Kurt(\Delta p)$ of dist. Q, small due to:
 - correlation of price changes for a given product (induced by b_i),
 - no time aggregation and very low measurement error.



Figure: Distribution of pooled price changes in a narrow category

Pooling all products "Non-durable household goods"

Estimated $Q(\cdot)$ and $q(\cdot)$, recovered $f(\cdot)$ and $\Lambda(\cdot)$







Estimating Kurtosis of underlying distribution

Recall:

 $\Delta p_{it} = b_i \Delta \tilde{p}_t \text{ for } i \in I \text{ and } t \in T(i)$ $\Delta \tilde{p}_t \sim Q \text{ and } b_i \perp \Delta \tilde{p}_t$

Then

$$Kurt(\Delta \tilde{p}_t) = \frac{Kurt(\Delta p_{it})}{1 + \operatorname{corr}(\Delta p_{it}^2, \Delta p_{is}^2)CV(\Delta p_{it}^2)CV(\Delta p_{is}^2)}$$

Autocorrelated square changes of product i inconsistent w/model

Autocorrelated square changes indicate heterogeneity.

Kurtosis Estimates for categories w/1000+ products

| Number Products | Number P. changes | $\hat{\mathbb{E}}(\Delta p_{it})$ | $\hat{\sigma}(\Delta p_{it})$ | Kurtosis Pooled | Kurtosis w/Unobs. Heterog. |
|--------------------|----------------------|-----------------------------------|-------------------------------|--------------------|----------------------------------|
| 3437 | 74464 | 0.002 | 0.342 | 3.367 (0.160) | 1.640 (0.065) |
| 3225 | 56527 | 0.002 | 0.329 | 3.807 (0.094) | 1.950 (0.047) |
| 2551 | 30343 | -0.001 | 0.246 | 3.520 (0.263) | 2.049 (0.155) |
| 1401 | 27321 | 0.002 | 0.344 | 2.923 (0.087) | 1.671 (0.051) |
| 1388 | 30111 | 0.003 | 0.309 | 3.579 (0.236) | 2.035 (0.116) |
| 1154 | 20995 | 0.007 | 0.309 | 3.467 (0.139) | 1.983 (0.049) |
| 1032 | 17724 | 0.002 | 0.261 | 3.314 (0.216) | 1.773 (0.128) |

Table: Bootstrapped standard errors in parenthesis

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GHF A and (duration) Survival Function S

• Let S(t) be survival function of duration of spell:

$$S(t) = \mathbb{E}\left[e^{-\int_0^t \Lambda(x(s))ds} \,|\, x(0) = 0\right]$$

assuming, to simplify, that $X = \infty$.

- Hazard $h(t) = -\frac{S'(t)}{S(t)}$ pr. Δp as function of **duration** t
- Hazard $h(\cdot)$, as function of duration is observable.
- Identification: Given S(t) (or h(t)) $\Longrightarrow \Lambda(x)$

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► For instance, using hazard rates *h*(*t*):

$$h(0) = \Lambda(0) \ge 0, \ \frac{\partial h(t)}{\partial t}|_{t=0} = \frac{\sigma^2}{2} \frac{\partial^2 \Lambda(x)}{\partial x^2}|_{x=0}$$
$$\frac{\partial^2 h(t)}{\partial t^2}|_{t=0} = \left(\frac{\sigma^2}{2}\right)^2 \frac{\partial^4 \Lambda(x)}{\partial x^4}|_{x=0},$$
$$\frac{\partial^3 h(t)}{\partial t^3}|_{t=0} = \left(\frac{\sigma^2}{2}\right)^3 \frac{\partial^6 \Lambda(x)}{\partial x^6}|_{x=0} - 4\left(\frac{\sigma^2}{2} \frac{\partial^2 \Lambda(x)}{\partial x^2}|_{x=0}\right)^2$$

GHF A and (duration) Survival Function *S* (*skip*)

Let S(t) be survival function of duration of spell:

$$S(t) = \mathbb{E}\left[e^{-\int_0^t \Lambda(x(s))ds} \,|\, x(0) = 0\right]$$

assuming, to simplify, that $X = \infty$.

- All the derivatives S(t) at t = 0 identify the level and all the even derivatives of Λ(x) at x = 0.
 Thus, if Λ is analytical, then it is identified by S.
- Using hazard rates:

$$h(0) = \Lambda(0) \ge 0, \ \frac{\partial h(t)}{\partial t}|_{t=0} = \frac{\sigma^2}{2} \frac{\partial^2 \Lambda(x)}{\partial x^2}|_{x=0}$$
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Measure of Price Changes independent of state

• Define *Calvo-ness* index $C \equiv \frac{\Lambda(0)}{N_a}$: fraction of price changes independent of the price gap x (state).

$$\mathcal{C} = q(0) \, rac{Var(\Delta p_{it})}{2 \, \mathbb{E}[|\Delta p_{it}|]}$$

for $t \neq s$.

- Density close to zero q(0) > 0 only if state not important.
- Can be adapted (see paper) w/unobserved heterogeneity.
- For Cavallo's data we estimate C ≈ 6%. Small values because small density of small price changes q(0).

Measure of Price Changes independent of state (skip)

• Define *Calvo-ness* index $C \equiv \frac{\Lambda(0)}{N_a}$, in words: fraction of price changes independent of the price gap (state).

Proposition:

$$\mathcal{C} = \mathcal{C}_{\textit{pooled}} \left(1 + \frac{\textit{Cov}(\textit{b}_i^{-1},\textit{b}_i^2)}{\mathbb{E}[\textit{b}_i^{-1}]\mathbb{E}[\textit{b}_i^2]} \right) < \mathcal{C}_{\textit{pooled}}$$

where the two components are given by

$$\mathcal{C}_{pooled} = \frac{q(0) \operatorname{Var}(\Delta p_{it})}{2 \operatorname{\mathbb{E}}[|\Delta p_{it}|]} \text{ and } 1 + \frac{\operatorname{Cov}(b_i^{-1}, b_i^2)}{\operatorname{\mathbb{E}}[b_i^{-1}] \operatorname{\mathbb{E}}[b_i^2]} = \frac{\operatorname{\mathbb{E}}[|\Delta p_{it}|^{-1}|\Delta p_{is}|^2]}{\operatorname{\mathbb{E}}[|\Delta p_{it}|^{-1}] \operatorname{\mathbb{E}}[|\Delta p_{it}|^2]}$$

for $t \neq s$.

For Cavallo's data we estimate $C \approx 6\%$. Small values because small density of small price changes q(0).

Cumulative Response to Aggregate Shock

- Once and for all monetary (cost) shock of size δ
- Effect of aggregate price level $\mathcal{P}(t; \delta)$ at horizon t
- Effect on output $Y(t; \delta) = (1/\epsilon) [\delta \mathcal{P}(t, \delta)]$ at horizon *t*.
- Cumulative IRF (CIRF) $\mathcal{M}(\delta) = \int_0^\infty Y(t, \delta) dt$
- Small shock $\mathcal{M}(\delta) \approx \mathcal{M}'(0)\delta$ since by definition $\mathcal{M}(0) = 0$.
- For models with no first order strategic complementarity.
 Keep same decision rules

• Show that
$$\mathcal{M}(\delta) = \frac{Kurt(\Delta p)}{6N_a}\delta + o(\delta^2)$$

Once and for all monetary shock at t = 0.



Impulse response of Price Level



Impulse response of output



Area under impulse response of output



Definition of CIR, cumulative response function (skip)

- Symmetry: suffices to consider effect until first Δp after shock δ .
- At t, a firm that has not adjusted contributes -x to output Y(t)
- m(x): contribution to CIR of firm starting w/gap x after shock δ .
- Let τ stopping time until |x| = X

$$m(x) = -\mathbb{E}\left[\int_0^\tau e^{-\int_0^t \Lambda(x(s))ds} x(t)dt \,|\, x(0) = x\right]$$

- ▶ Just before the shock, firms gap x are distributed as $f(\cdot)$.
- Since shock moves each x to $x \delta$, CIR is given by

$$\mathcal{M}(\delta) = \int_{-X}^{X} m(x) f(x+\delta) dx = \delta \int_{-X}^{X} m(x) f'(x) dx + o(\delta^2)$$

Sufficient Statistic for CIR, first result

► Let ∧ be ANY Generalized Hazard Function.

Then the cumulative impulse response (CIR) for a small monetary shock:

$$\mathcal{M}(\delta) = \mathcal{M}'(\mathbf{0})\delta + o(\delta^2) = \frac{Kurt(\Delta p)}{6N_a}\delta + o(\delta^2)$$

The approximation is accurate, since $\mathcal{M}''(0) = 0$.

- Kurt(Δp) depends on shape of $\Lambda(\cdot)$, not its level
- multiplying $\Lambda(\cdot) \& \sigma^2$ by constant \implies same dist. $q \& Kurt(\Delta p)$.
- multiplying $\Lambda(\cdot)$ and σ^2 by constant scales N_a by same constant.

Sufficient Statistic for CIR, bounds, second result

Recall

$$\mathcal{M}(\delta) = rac{Kurt(\Delta p)}{6N_a}\delta + o(\delta^2)$$

- Only weakly increasing A can be rationalized as random menu cost or optimal adjustment intensity model.
- Among the weakly increasing Λ the inverted-L hazard Λ (i.e. Golosov and Lucas) has the smallest Kurt(Δp) = 1.
- Among the weakly increasing Λ the <u>constant</u> hazard Λ (i.e. Calvo) has the largest Kurt(Δp) = 6.
- Decreasing Λ have $Kurt(\Delta p) > 6$, and can be arbitrarily large.
- These results imply that Kurt(Δp) measure positive selection for price increases after a positive aggregate shock.

Scope & limitations of $Kurt(\Delta p)/(6 N_a)$ skip if short in time as a Cumulative Impulse Response (*CIR*) sufficient statistic

- Also hold on:
 - Calvo⁺ + Multiproduct n ≥ 1 products, (Alvarez-Lebihan-Lippi AER 16), common case Calvo⁺ w/n = 1.
 - General Rational inattentiveness model (Reis 06), (Alvarez-Lippi-Paciello, RES 15) common case: pure Calvo.
 - Insensitive, up to first order, to add:
 - steady state inflation
 - asymmetry in objective function
- Does NOT hold:
 - Large inflation (Sheshinsky-Weiss, Blanco-Bailey)
 - Firm do not close gap (Alvarez-Lippi AEJ 19) Temporary price changes.
 - Stochastic volatility.
 - Strategic Interactions (?).

Beyond CIR: Impulse Response $Y(t; \delta) = \delta - \mathcal{P}(t, \delta)$

- Use Eigenvalue-Eigenfunction to solve for $Y(t; \delta)$ and $\mathcal{P}(t, \delta)$.
- Caballero-Engel meet Schrodinger-Dirac:
- Quadratic A has the same eigenfunctions-eigenvalues as famous "Quantum-Oscillator".
- In general, Schrodinger Eqn for positive symmetric potential.
- Use this result to compare with other proposed sufficient statistic: Flexibility Index.
- Flexibility Index $\mathcal{F} \equiv \lim_{t \to 0, \delta \to 0} \frac{\partial}{\partial_t} Y(t, \delta)$
- initial slope of the IRF with respect to time, at a small shock.
- Flexibility Index F does not give same ordering as M CIR.
- Construct examples with same \mathcal{F} & are very different CIR \mathcal{M} .
- Even effect for $Y(t; \delta)$ for small t can be quite different.

▶ IRF of price level, recall $Y(t; \delta) = \delta - \mathcal{P}(t, \delta)$:

$$\mathcal{P}(t,\delta) = \Omega(\delta) + \int_0^t \omega(s,\delta) \, ds$$

• $\omega(t, \delta) = \frac{\partial}{\partial t} \mathcal{P}(t, \delta)$ flow at *t*, and $\Omega(\delta)$ initial jump at t = 0:

$$\omega(t,\delta) = -\int_{-X}^{X} x \Lambda(x) f(x,t) dx + X \sigma^2 \left[f'(-X,t) - f'(X,t) \right]$$
$$\Omega(\delta) = \int_{-X}^{-X+\delta} (-x+\delta) f(x,0) dx$$

• f(x, t) solves KFE with initial condition $f(x, 0) = f(x + \delta)$ does not have to be steady state.

• Flexibility Index
$$\mathcal{F} \equiv \frac{\partial}{\partial \delta} \omega(\mathbf{0}, \delta)|_{\delta = \mathbf{0}}$$
,

• If $X < \infty \implies$ Flexibility Index $\mathcal{F} = \infty$

skip

Simplify arguments by using X = ∞,
 (Flexibility index is always infinite with barriers)

• IRF of price level, recall $Y(t; \delta) = \delta - \mathcal{P}(t, \delta)$:

$$\mathcal{P}(t,\delta) = \int_0^t \omega(s,\delta) \, ds$$

• $\omega(t,\delta) = \frac{\partial}{\partial t} \mathcal{P}(t,\delta)$ flow at t,

$$\omega(t,\delta) = -\int_{-X}^{X} x \Lambda(x) f(x,t) dx$$

- f(x, t) solves KFE with initial condition $f(x, 0) = f(x + \delta)$
- Flexibility Index $\mathcal{F} \equiv \frac{\partial}{\partial \delta} \omega(\mathbf{0}, \delta)|_{\delta = \mathbf{0}}$,

▶ \mathcal{F} is easy to compute, only requires f(x, 0) and gives:

$$\mathcal{F} = N_a \left(1 + \int_{-\infty}^{\infty} \frac{x \Lambda'(x)}{\Lambda(x)} q(x) dx \right)$$

skip, if short of time

• Flexibility Index
$$\mathcal{F} \equiv \frac{\partial}{\partial \delta} \omega(\mathbf{0}, \delta)|_{\delta = \mathbf{0}}$$
,

Advantages of \mathcal{F} : is easy to compute, only requires f(x, 0):

$$\mathcal{F} = N_a \left(1 + \int_{-\infty}^{\infty} \frac{x \Lambda'(x)}{\Lambda(x)} q(x) dx \right)$$

 Disadvantages of Flexibility Index: By design, it only measures effect at very short term.

- Example: $\Lambda(x) = \Lambda(0) + \kappa x^{\nu}$ so that:
 - Same frequency of adjustment $N_a = 1$
 - Same Flexibility Index F = 3
 - Different Cumulative Impulse Response M, i.e. different Kurtosis.
- Literature "misinterpret" idea of sufficient statistic.

Figure: Values of *Kurt*(Δp) or CIR relative to quadratic $\Lambda(x)$ Each case (a dot), different parameters ν for $\Lambda(x) = \Lambda(0) + \kappa x^{\nu}$ All cases have values $\Lambda(0), \kappa$ so that they have the same $\mathcal{F} = 3$ and $N_a = 1$



Figure: Impulse Responses for two cases of power plus $\Lambda(x) = \Lambda(0) + \kappa x^{\nu}$ Both cases with same \mathcal{F} and N_a . Same slope at zero Y'(0), but IRF Y(t) different even in short run, since expected time to adjustment $N_a = 1$



Compute Entire Impulse Response function (skip)

- Consider operator ^{*d*²}/₂ ^{*d*²}/_{*dx*²} − Λ(*x*) on functions *g* on [−*X*, *X*], Direchlet boundary *g*(−*X*) = *g*(*X*) = 0.
- If $X = \infty$ assume $\Lambda(x) \to \infty$, so operator is compact.
- Impulse response on shift δ to invariant f for functiong is

$$m{Y}(t) = \delta \sum_{j=1}^{\infty} m{e}^{\lambda_j t} \langle arphi_j, m{g}
angle \langle arphi_j, , f'(m{x})
angle$$

•
$$\varphi_j$$
 eigenfunctions, λ_j eigenvalues.

A lot is known about eigenfunctions and eigenvalues.

Conclusions

- Analyze popular model in the literature, Caballero and Engel, Dotsey-King-Wollman, Woodford,
- Models: random-menu cost G or optimal adjustment intensity c
- Any weakly increasing A can be rationalized
- Sufficient statistic for CIR of monetary shock $\frac{Kurt(\Delta p)}{6N_2}$
- Span from GL (smallest effect) to Calvo (largest effect)
- Characterization: $G(\text{ or } c) \iff \Lambda \iff f \iff q$
- Characterization: $S \iff \Lambda$, where S survival function (duration).
- ▶ Quadratic Λ , used in literature: small effects, \approx Cavallo's data
- Define & estimate small % of price changes independent of state.
- Eigenvalues-Eigenfunctions used to computing entire IRF
- Examples: Flexibility index is constant, but CIR changes a lot.