

CAUSAL INFERENCE FOR ASSET PRICING

Valentin Haddad Zhiguo He Paul Huebner Peter Kondor Erik Loualiche

UCLA, Stanford, SSE, LSE, Minnesota

June 2025

CAUSAL INFERENCE FOR ASSET PRICING

Causal inference methods

- e.g. use IV/diff-in-diff to learn about investors' portfolio choice or equilibrium asset prices
 - If the stock price of Tesla drops by 1%, how do you change your position?
 - If a group of investors starts buying GameStop, how does its price change?
- *natural experiments + inference without taking a strong stand on mechanism*

CAUSAL INFERENCE FOR ASSET PRICING

Causal inference methods

- e.g. use IV/diff-in-diff to learn about investors' portfolio choice or equilibrium asset prices
 - If the stock price of Tesla drops by 1%, how do you change your position?
 - If a group of investors starts buying GameStop, how does its price change?
- *natural experiments + inference without taking a strong stand on mechanism*

“This is not how we do asset pricing”

CAUSAL INFERENCE FOR ASSET PRICING

Causal inference methods

- e.g. use IV/diff-in-diff to learn about investors' portfolio choice or equilibrium asset prices
 - If the stock price of Tesla drops by 1%, how do you change your position?
 - If a group of investors starts buying GameStop, how does its price change?
- *natural experiments + inference without taking a strong stand on mechanism*

Traditional asset pricing empirical methods

- Euler equations, factor models, Epstein-Zin preferences, ...
- *equilibrium relations + fully specified models*

CAUSAL INFERENCE FOR ASSET PRICING

Causal inference methods

- e.g. use IV/diff-in-diff to learn about investors' portfolio choice or equilibrium asset prices
 - If the stock price of Tesla drops by 1%, how do you change your position?
 - If a group of investors starts buying GameStop, how does its price change?
- *natural experiments + inference without taking a strong stand on mechanism*

Traditional asset pricing empirical methods

- Euler equations, factor models, Epstein-Zin preferences, ...
- *equilibrium relations + fully specified models*

→ **This paper:** a causal inference framework that is compatible with finance ideas

AN EXAMPLE

- Hengjie has detailed data on corporate bond holdings of CalPERS and bond prices
- *How would portfolio positions change if, all else equal, bond prices move?*
 - demand curve of CalPERS

AN EXAMPLE

- Hengjie has detailed data on corporate bond holdings of CalPERS and bond prices
- *How would portfolio positions change if, all else equal, bond prices move?*
 - demand curve of CalPERS
- OLS is a bad idea:
 - prices affect CalPERS demand
 - CalPERS (and many others) demand affect prices

AN EXAMPLE

- Hengjie has detailed data on corporate bond holdings of CalPERS and bond prices
- *How would portfolio positions change if, all else equal, bond prices move?*
 - demand curve of CalPERS
- OLS is a bad idea:
 - prices affect CalPERS demand
 - CalPERS (and many others) demand affect prices
- **Natural experiment:** the Fed decides to do a one-off intervention buying random corporate bonds

HENGJIE'S DILEMMA

Canonical causal inference: IV with $Z_i = \text{Fed purchases of bond } i$

$$\Delta D_i = \hat{\mathcal{E}} \Delta P_i + \theta' X_i + e_i$$

$$\Delta P_i = \lambda Z_i + \eta' X_i + u_i$$

HENGJIE'S DILEMMA

Canonical causal inference: IV with $Z_i = \text{Fed purchases of bond } i$

$$\Delta D_i = \hat{\mathcal{E}} \Delta P_i + \theta' X_i + e_i$$

$$\Delta P_i = \lambda Z_i + \eta' X_i + u_i$$

Finance: holdings decided as a portfolio

- When price of a green bond increases, CalPERS sells some of it ... and replace by investing disproportionately more in other green bonds than brown bonds

$$\Delta D = \underset{\text{matrix}}{\mathcal{E}} \Delta P + \epsilon$$

HENGJIE'S DILEMMA

Canonical causal inference: IV with Z_i = Fed purchases of bond i

$$\Delta D_i = \hat{\mathcal{E}} \Delta P_i + \theta' X_i + e_i$$

$$\Delta P_i = \lambda Z_i + \eta' X_i + u_i$$

Finance: holdings decided as a portfolio

- When price of a green bond increases, CalPERS sells some of it ... and replace by investing disproportionately more in other green bonds than brown bonds

$$\Delta D = \underset{\text{matrix}}{\mathcal{E}} \Delta P + \epsilon$$

→ Challenge of demand estimation with many goods

- SUTVA is violated
- All other prices are omitted variables ... too many to instrument them all

THIS PAPER: ASSUMPTION

1. An elementary assumption about demand: homogeneous substitution conditional on observables

- When CalPERS substitutes, does so differentially for bonds with different observables (e.g. greenness, duration), but does not distinguish between bonds with the same observables
- Markowitz finance: factor structure of covariance matrix
- Many others: targeting of portfolio level targets (e.g. regulatory scores), logit, ...
- Empirical design, supporting evidence, ...

THIS PAPER: IDENTIFICATION

2. **Cross-sectional causal inference** identifies the **relative elasticity** ($\hat{\mathcal{E}}$):

- How does the relative demand for two bonds with the same greenness and duration respond to a change in their relative price?
- Difference between own-price and cross-price elasticity

THIS PAPER: IDENTIFICATION

2. **Cross-sectional causal inference** identifies the **relative elasticity** ($\hat{\mathcal{E}}$):

- How does the relative demand for two bonds with the same greenness and duration respond to a change in their relative price?
- Difference between own-price and cross-price elasticity
- *Partial identification*: no information about substitution patterns

THIS PAPER: IDENTIFICATION

2. **Cross-sectional causal inference** identifies the **relative elasticity** ($\hat{\mathcal{E}}$):

- How does the relative demand for two bonds with the same greenness and duration respond to a change in their relative price?
- Difference between own-price and cross-price elasticity
- *Partial identification*: no information about substitution patterns

3. **A small set of time series regressions** identifies **substitution**

- Substitution = meso and macro elasticity
 - Meso: How does the demand for green bonds relative to brown bonds respond to the price of the green-minus-brown portfolio and the market portfolio?
 - Macro: How does the demand for bonds responds to the price of the market portfolio and the green-minus-brown portfolio?
- Needs simultaneous instruments over time for the price of all portfolios

RELATED LITERATURE

■ Asset pricing using causal inference methods

- Shleifer (1986); Harris, Gurel (1986); Chang, Hong, Liskovich (2014); Pavlova, Sikorskaya (2023); Greenwood, Sammon (2024); Gompers, Metrick (2001); Coval, Stafford (2007); Lou (2012); Ben-David, Li, Rossi, Song (2022); Hartzmark, Solomon (2022); Krishnamurthy, Vissing-Jorgensen (2011); Haddad, Moreira, Muir (2021, 2025); Selgrad (2024); Du, Tepper, Verdelhan (2018); Greenwood, Vissing-Jorgensen (2018); Haddad, Muir (2021); Chen, Chen, He, Liu, Xie (2023); ...

■ Structural approach and demand systems

- *Koijen, Yogo (2019, 2024); Koijen, Richmond, Yogo (2024);* Haddad, Huebner, Loualiche (2024); van der Beck (2024); Lu, Wu (2023); Gabaix, Koijen (2024); Bretscher, Schmid, Sen, Sharma (2024); Jansen, Li, Schmid (2024); Fang (2023); Fang, Xiao (2024); ...
- Li, Lin (2024); Chaudhary, Fu, Li (2023); Aghaee (2024); An, Huber (2025); Peng, Wang (2023); Fuchs, Fukuda, Neuhaus (2024); ...

■ What we don't do:

- Strategic responses (Haddad, Huebner, Loualiche, 2024)
- Dynamics (Greenwood, Hanson, Liao, 2018; Huebner, 2024; Gabaix, Koijen, 2024; He, Kondor, Li)
- State-contingent shocks (Haddad, Moreira, Muir, 2025)

■ Spillovers/substitution outside asset pricing:

- *Berry, Levinsohn, Pakes (1995),* Berg, Reisinger, Streitz (2021); Chodorow-Reich, Nenov, Simsek (2021); Guren, McKay, Nakamura, Steinsson (2021), Huber (2023); Wolf (2023); ...

OUTLINE

- 1 HOMOGENEOUS SUBSTITUTION CONDITIONAL ON OBSERVABLES
- 2 CROSS-SECTIONAL CAUSAL INFERENCE
- 3 ESTIMATING SUBSTITUTION
- 4 ESTIMATING MULTIPLIERS

AN ASSUMPTION
FOR DEMAND IN ASSET PRICING

THE DEMAND FUNCTION FOR ASSETS

- Investor chooses portfolio ... taking prices are given: $D(P, \dots)$
 - Similar structure with market power or learning from prices: post a demand curve

THE DEMAND FUNCTION FOR ASSETS

- Investor chooses portfolio ... taking prices are given: $D(P, \dots)$
 - Similar structure with market power or learning from prices: post a demand curve
- In changes, linearize:

$$\Delta D = \underset{\text{elasticity}}{\mathcal{E}} \Delta P + \underset{\text{shifts}}{\epsilon}$$

- could be logs, levels, portfolio weights, yields: choose to give regularity to \mathcal{E}

THE DEMAND FUNCTION FOR ASSETS

- Investor chooses portfolio ... taking prices are given: $D(P, \dots)$
 - Similar structure with market power or learning from prices: post a demand curve
- In changes, linearize:

$$\Delta D = \underset{\text{elasticity}}{\mathcal{E}} \Delta P + \underset{\text{shifts}}{\epsilon}$$

- could be logs, levels, portfolio weights, yields: choose to give regularity to \mathcal{E}
- Markowitz: $D = \frac{1}{\gamma} \Sigma^{-1} (\mu - P) \Rightarrow \mathcal{E} = \frac{1}{\gamma} \Sigma^{-1}$

THE DEMAND FUNCTION FOR ASSETS

- Investor chooses portfolio ... taking prices are given: $D(P, \dots)$
 - Similar structure with market power or learning from prices: post a demand curve
- In changes, linearize:

$$\Delta D = \underset{\text{elasticity}}{\mathcal{E}} \Delta P + \underset{\text{shifts}}{\epsilon}$$

- could be logs, levels, portfolio weights, yields: choose to give regularity to \mathcal{E}
- Markowitz: $D = \frac{1}{\gamma} \Sigma^{-1} (\mu - P) \Rightarrow \mathcal{E} = \frac{1}{\gamma} \Sigma^{-1}$
- *Multipliers* or price impact: effect of demand shifts on equilibrium prices $\mathcal{M} = \mathcal{E}_{agg}^{-1}$

THE DEMAND FUNCTION FOR ASSETS

- Investor chooses portfolio ... taking prices are given: $D(P, \dots)$
 - Similar structure with market power or learning from prices: post a demand curve
- In changes, linearize:

$$\Delta D = \underset{\text{elasticity}}{\mathcal{E}} \Delta P + \underset{\text{shifts}}{\epsilon}$$

- could be logs, levels, portfolio weights, yields: choose to give regularity to \mathcal{E}
- Markowitz: $D = \frac{1}{\gamma} \Sigma^{-1} (\mu - P) \Rightarrow \mathcal{E} = \frac{1}{\gamma} \Sigma^{-1}$
- *Multipliers* or price impact: effect of demand shifts on equilibrium prices $\mathcal{M} = \mathcal{E}_{agg}^{-1}$

This talk: what if you want to figure \mathcal{E} out from data without assuming much?

AN ELEMENTARY ASSUMPTION

A1. **Homogeneous substitution conditional on observables**

→ Any pair of assets in the estimation sample \mathcal{S} with the same observables shares the same cross-price elasticity with respect to any third asset, within or outside the estimation sample:

$$\boxed{\mathcal{E}_{il} = \mathcal{E}_{jl} \text{ if } X_i = X_j} \quad \text{for all } i, j \in \mathcal{S}, \text{ and } l \neq i, j,$$

- When substituting, the investor differentiates with respect to X only

AN ELEMENTARY ASSUMPTION

A1. **Homogeneous substitution conditional on observables**

→ Any pair of assets in the estimation sample \mathcal{S} with the same observables shares the same cross-price elasticity with respect to any third asset, within or outside the estimation sample:

$$\boxed{\mathcal{E}_{il} = \mathcal{E}_{jl} \text{ if } X_i = X_j} \quad \text{for all } i, j \in \mathcal{S}, \text{ and } l \neq i, j,$$

- When substituting, the investor differentiates with respect to X only
- X_i : $K \times 1$ vector of observables for asset i .
- Add linearity $\mathcal{E}_{il} = \mathcal{E}_{\text{cross}}(X_i, X_l) = X_i' \underbrace{\mathcal{E}_X}_{K \times K} X_l$

AN ELEMENTARY ASSUMPTION

A1. **Homogeneous substitution conditional on observables**

→ Any pair of assets in the estimation sample \mathcal{S} with the same observables shares the same cross-price elasticity with respect to any third asset, within or outside the estimation sample:

$$\boxed{\mathcal{E}_{il} = \mathcal{E}_{jl} \text{ if } X_i = X_j} \quad \text{for all } i, j \in \mathcal{S}, \text{ and } l \neq i, j,$$

- When substituting, the investor differentiates with respect to X only
- X_i : $K \times 1$ vector of observables for asset i .
- Add linearity $\mathcal{E}_{il} = \mathcal{E}_{\text{cross}}(X_i, X_l) = X_i' \underbrace{\mathcal{E}_X}_{K \times K} X_l$
- Can apply everywhere, or just to a sample of assets \mathcal{S}

REGULARIZING A BIT MORE

A2. Constant relative elasticity

→ *Assets in the estimation sample have the same value of relative elasticity $\mathcal{E}_{relative}$ with respect to other assets with the same characteristics:*

$$\boxed{\mathcal{E}_{ii} - \mathcal{E}_{ji} = \mathcal{E}_{relative} \text{ if } X_i = X_j} \quad \text{for all } i, j \in \mathcal{S}$$

- How does the relative demand for two assets with the same observables respond to a change in their relative price?

REGULARIZING A BIT MORE

A2. Constant relative elasticity

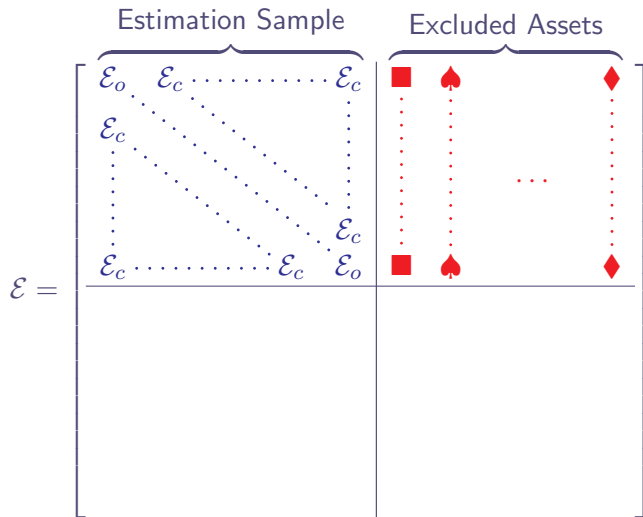
→ *Assets in the estimation sample have the same value of relative elasticity $\mathcal{E}_{relative}$ with respect to other assets with the same characteristics:*

$$\boxed{\mathcal{E}_{ii} - \mathcal{E}_{ji} = \mathcal{E}_{relative} \text{ if } X_i = X_j} \quad \text{for all } i, j \in \mathcal{S}$$

- How does the relative demand for two assets with the same observables respond to a change in their relative price?
- Similar local behavior across assets → homogeneous treatment effect
- Can relax a lot for cross-sectional results (function of characteristics, LATE)

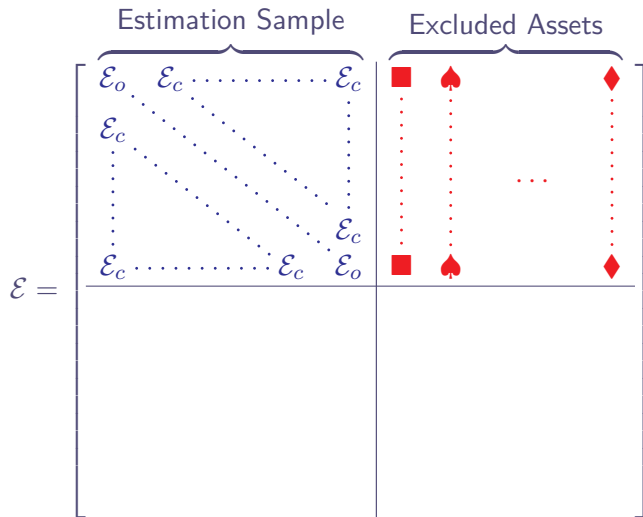
USING THE ASSUMPTIONS: LOCAL EXPERIMENTS

- With few close assets, could ignore observables and assume full homogeneity
 - Demand for 10-yr bonds of Ford and GM responds in same way to price of 5-year First Solar bond



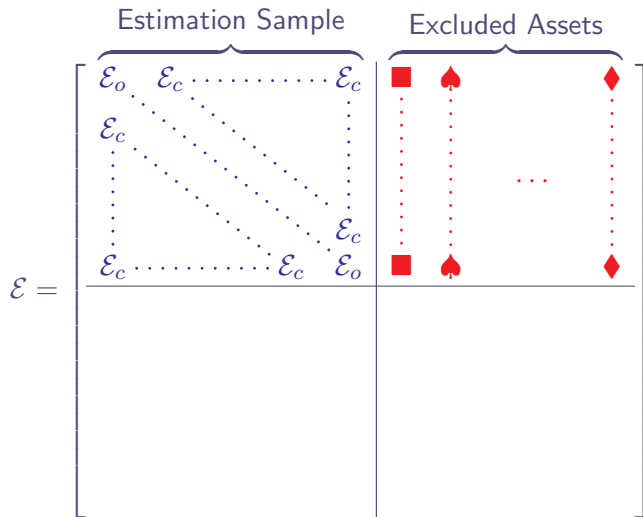
USING THE ASSUMPTIONS: LOCAL EXPERIMENTS

- With few close assets, could ignore observables and assume full homogeneity
 - Demand for 10-yr bonds of Ford and GM responds in same way to price of 5-year First Solar bond
- Risk-based models: assets have common variance and covariance + *identical covariance with outside assets*



USING THE ASSUMPTIONS: LOCAL EXPERIMENTS

- With few close assets, could ignore observables and assume full homogeneity
 - Demand for 10-yr bonds of Ford and GM responds in same way to price of 5-year First Solar bond
- Risk-based models: assets have common variance and covariance + *identical covariance with outside assets*
- Diagnostic: balance between treated (high Z_i) and control (low Z_i) on covariance with broad factors



USING THE ASSUMPTIONS: RICH CROSS-SECTIONS

Key question: What does the investor consider when substituting between assets?

- *Investor manages <aggregate statistic> hence substitution depends on <individual asset contribution>*

USING THE ASSUMPTIONS: RICH CROSS-SECTIONS

Key question: What does the investor consider when substituting between assets?

- *Investor manages <aggregate statistic> hence substitution depends on <individual asset contribution>*
- *Broad categories: X_i are group dummies*

USING THE ASSUMPTIONS: RICH CROSS-SECTIONS

Key question: What does the investor consider when substituting between assets?

- *Investor manages <aggregate statistic> hence substitution depends on <individual asset contribution>*
- *Broad categories: X_i are group dummies*
- *Risk based motives: care about portfolio-level factor exposure, so X_i are factor loadings or characteristics that proxy for them*

USING THE ASSUMPTIONS: RICH CROSS-SECTIONS

Key question: What does the investor consider when substituting between assets?

- *Investor manages <aggregate statistic> hence substitution depends on <individual asset contribution>*
- *Broad categories:* X_i are group dummies
- *Risk based motives:* care about portfolio-level factor exposure, so X_i are factor loadings or characteristics that proxy for them
- *Non-risk motives:* X_i is asset weight in this objective
 - Binding constraints (e.g. leverage)
 - Manages a regulatory score (e.g. capital ratio,...)
 - Stakeholders pressure (greenness, ...)

$$\begin{aligned} \max_D \quad & D'(M - P) - \frac{\gamma}{2} D' \Sigma D - \frac{\kappa}{2} \left(D' X^{(1)} \right)^2 \\ \text{such that} \quad & D' X^{(2)} \leq \Theta \end{aligned}$$

CROSS-SECTIONAL CAUSAL INFERENCE

CROSS-SECTIONAL IDENTIFICATION

Back to Hengjie ...

CROSS-SECTIONAL IDENTIFICATION

Back to Hengjie ...

- He knows true model is matrix with CalPERS caring about greenness and duration (X)

$$\Delta D = \mathcal{E} \Delta P + \epsilon$$

- He runs the regression:

$$\Delta D_i = \hat{\mathcal{E}} \Delta P_i + \theta' X_i + e_i$$

$$\Delta P_i = \lambda Z_i + \eta' X_i + u_i$$

CROSS-SECTIONAL IDENTIFICATION

Back to Hengjie ...

- He knows true model is matrix with CalPERS caring about greenness and duration (X)

$$\Delta D = \mathcal{E} \Delta P + \epsilon$$

- He runs the regression:

$$\Delta D_i = \hat{\mathcal{E}} \Delta P_i + \theta' X_i + e_i$$

$$\Delta P_i = \lambda Z_i + \eta' X_i + u_i$$

Proposition 1. Under A1, A2, and the usual exclusion and relevance restrictions, the IV estimator identifies the **relative elasticity** $\hat{\mathcal{E}} = \mathcal{E}_{relative} = \mathcal{E}_{ii} - \mathcal{E}_{ji}$.

CROSS-SECTIONAL IDENTIFICATION

Back to Hengjie ...

- He knows true model is matrix with CalPERS caring about greenness and duration (X)

$$\Delta D = \mathcal{E} \Delta P + \epsilon$$

- He runs the regression:

$$\Delta D_i = \hat{\mathcal{E}} \Delta P_i + \theta' X_i + e_i$$

$$\Delta P_i = \lambda Z_i + \eta' X_i + u_i$$

Proposition 1. Under A1, A2, and the **usual exclusion and relevance restrictions**, the IV estimator identifies the **relative elasticity** $\hat{\mathcal{E}} = \mathcal{E}_{relative} = \mathcal{E}_{ii} - \mathcal{E}_{ji}$.

MECHANICS FOR THE SIMPLE CASE

- Take 2 assets with same characteristics, $X_1 = X_2$

First difference

$$\Delta D_1 = \varepsilon_{11}\Delta P_1 + \varepsilon_{12}\Delta P_2 + \sum_{k \geq 3} \varepsilon_{1k}\Delta P_k$$

$$\Delta D_2 = \varepsilon_{22}\Delta P_2 + \varepsilon_{21}\Delta P_1 + \sum_{k \geq 3} \varepsilon_{2k}\Delta P_k$$

Second difference

MECHANICS FOR THE SIMPLE CASE

- Take 2 assets with same characteristics, $X_1 = X_2$

First difference

$$\begin{aligned}\Delta D_1 &= \mathcal{E}_{11}\Delta P_1 + \mathcal{E}_{12}\Delta P_2 + \sum_{k \geq 3} \mathcal{E}_{1k}\Delta P_k \\ &= \mathcal{E}_{11}\Delta P_1 + \mathcal{E}_{\text{cross}}(X_1, X_2)\Delta P_2 + \sum_{k \geq 3} \mathcal{E}_{\text{cross}}(X_1, X_k)\Delta P_k \\ \Delta D_2 &= \mathcal{E}_{22}\Delta P_2 + \mathcal{E}_{21}\Delta P_1 + \sum_{k \geq 3} \mathcal{E}_{2k}\Delta P_k\end{aligned}$$

Second difference

MECHANICS FOR THE SIMPLE CASE

- Take 2 assets with same characteristics, $X_1 = X_2$

First difference

$$\begin{aligned}\Delta D_1 &= \varepsilon_{11}\Delta P_1 + \varepsilon_{12}\Delta P_2 + \sum_{k \geq 3} \varepsilon_{1k}\Delta P_k \\ &= \varepsilon_{11}\Delta P_1 + \mathcal{E}_{\text{cross}}(X_2, X_2)\Delta P_2 + \sum_{k \geq 3} \mathcal{E}_{\text{cross}}(X_1, X_k)\Delta P_k \\ \Delta D_2 &= \varepsilon_{22}\Delta P_2 + \varepsilon_{21}\Delta P_1 + \sum_{k \geq 3} \varepsilon_{2k}\Delta P_k\end{aligned}$$

Second difference

MECHANICS FOR THE SIMPLE CASE

- Take 2 assets with same characteristics, $X_1 = X_2$

First difference

$$\begin{aligned}\Delta D_1 &= \mathcal{E}_{11}\Delta P_1 + \mathcal{E}_{12}\Delta P_2 + \sum_{k \geq 3} \mathcal{E}_{1k}\Delta P_k \\ &= \mathcal{E}_{11}\Delta P_1 + \mathcal{E}_{\text{cross}}(X_2, X_2)\Delta P_2 + \sum_{k \geq 3} \mathcal{E}_{\text{cross}}(X_1, X_k)\Delta P_k \\ \Delta D_2 &= \mathcal{E}_{22}\Delta P_2 + \mathcal{E}_{21}\Delta P_1 + \sum_{k \geq 3} \mathcal{E}_{2k}\Delta P_k \\ &= \mathcal{E}_{22}\Delta P_2 + \mathcal{E}_{\text{cross}}(X_1, X_1)\Delta P_1 + \sum_{k \geq 3} \mathcal{E}_{\text{cross}}(X_1, X_k)\Delta P_k\end{aligned}$$

Second difference

MECHANICS FOR THE SIMPLE CASE

- Take 2 assets with same characteristics, $X_1 = X_2$

First difference

$$\begin{aligned}\Delta D_1 &= \mathcal{E}_{11}\Delta P_1 + \mathcal{E}_{12}\Delta P_2 + \sum_{k \geq 3} \mathcal{E}_{1k}\Delta P_k \\ &= \mathcal{E}_{11}\Delta P_1 + \mathcal{E}_{\text{cross}}(X_2, X_2)\Delta P_2 + \sum_{k \geq 3} \mathcal{E}_{\text{cross}}(X_1, X_k)\Delta P_k\end{aligned}$$

$$\begin{aligned}\Delta D_2 &= \mathcal{E}_{22}\Delta P_2 + \mathcal{E}_{21}\Delta P_1 + \sum_{k \geq 3} \mathcal{E}_{2k}\Delta P_k \\ &= \mathcal{E}_{22}\Delta P_2 + \mathcal{E}_{\text{cross}}(X_1, X_1)\Delta P_1 + \sum_{k \geq 3} \mathcal{E}_{\text{cross}}(X_1, X_k)\Delta P_k\end{aligned}$$

Second difference

$$\Delta D_1 - \Delta D_2 = \underbrace{(\mathcal{E}_{11} - \mathcal{E}_{\text{cross}}(X_1, X_1))}_{\text{}} \Delta P_1 - \underbrace{(\mathcal{E}_{22} - \mathcal{E}_{\text{cross}}(X_2, X_2))}_{\text{}} \Delta P_2$$

MECHANICS FOR THE SIMPLE CASE

- Take 2 assets with same characteristics, $X_1 = X_2$

First difference

$$\begin{aligned}\Delta D_1 &= \varepsilon_{11}\Delta P_1 + \varepsilon_{12}\Delta P_2 + \sum_{k \geq 3} \varepsilon_{1k}\Delta P_k \\ &= \varepsilon_{11}\Delta P_1 + \mathcal{E}_{\text{cross}}(X_2, X_2)\Delta P_2 + \sum_{k \geq 3} \mathcal{E}_{\text{cross}}(X_1, X_k)\Delta P_k\end{aligned}$$

$$\begin{aligned}\Delta D_2 &= \varepsilon_{22}\Delta P_2 + \varepsilon_{21}\Delta P_1 + \sum_{k \geq 3} \varepsilon_{2k}\Delta P_k \\ &= \varepsilon_{22}\Delta P_2 + \mathcal{E}_{\text{cross}}(X_1, X_1)\Delta P_1 + \sum_{k \geq 3} \mathcal{E}_{\text{cross}}(X_1, X_k)\Delta P_k\end{aligned}$$

Second difference

$$\begin{aligned}\Delta D_1 - \Delta D_2 &= \underbrace{(\varepsilon_{11} - \varepsilon_{\text{cross}}(X_1, X_1))}_{\hat{\mathcal{E}}} \Delta P_1 - \underbrace{(\varepsilon_{22} - \varepsilon_{\text{cross}}(X_2, X_2))}_{\hat{\mathcal{E}}} \Delta P_2 \\ &= \hat{\mathcal{E}}(\Delta P_1 - \Delta P_2)\end{aligned}$$

WHAT ABOUT EQUILIBRIUM?

- In the data, changes in price are always in equilibrium, so no notion of “separately moving each price”
 - If the Fed doesn't buy a bond, its price might move anyways because it bought its substitutes

WHAT ABOUT EQUILIBRIUM?

- In the data, changes in price are always in equilibrium, so no notion of “separately moving each price”
 - If the Fed doesn't buy a bond, its price might move anyways because it bought its substitutes
- **Never an issue for exogeneity**
 - Exogeneity is “Fed buying a bond or not” uncorrelated to demand shifts of CalPERS
 - $Z_i \perp \epsilon_i | X_i$

WHAT ABOUT EQUILIBRIUM?

- In the data, changes in price are always in equilibrium, so no notion of “separately moving each price”
 - If the Fed doesn't buy a bond, its price might move anyways because it bought its substitutes
- **Never an issue for exogeneity**
 - Exogeneity is “Fed buying a bond or not” uncorrelated to demand shifts of CalPERS
 - $Z_i \perp \epsilon_i | X_i$
- If equilibrium is such that the two prices cannot deviate *at all* from each other, relevance might fail
 - You can assess this empirically!

THE MISSING PIECE: SUBSTITUTION

- Key step: control for characteristics θ absorbs substitution from other assets

$$\sum_{k \geq 3} \mathcal{E}_{\text{cross}}(X_1, X_k) \Delta P_k = X_1' \underbrace{\mathcal{E}_X X \Delta P}_{\text{constant in data}}$$

THE MISSING PIECE: SUBSTITUTION

- Key step: control for characteristics θ absorbs substitution from other assets

$$\sum_{k \geq 3} \mathcal{E}_{\text{cross}}(X_1, X_k) \Delta P_k = X_1' \underbrace{\mathcal{E}_X X \Delta P}_{\text{constant in data}}$$

Absorbing substitution \neq Estimating substitution

- "Missing intercept and coefficients" problem: doesn't know how θ would change with different prices

WHAT ABOUT LOGIT?

- Koijen Yogo (1999), Koijen Richmond Yogo (2024):

$$\textit{“log utility + factor model = logit”}$$

- Logit accounts for substitution, so nothing missing

WHAT ABOUT LOGIT?

- Koijen Yogo (1999), Koijen Richmond Yogo (2024):

$$\textit{“log utility + factor model = logit”}$$

- Logit accounts for substitution, so nothing missing
- **An arbitrary factor model is not equivalent to logit**
 - Logit: when the price of any bond increases, CalPERS replaces it proportionally to its existing portfolio
 - Factor model: when the price of a bond increases, CalPERS replaces it disproportionately with bonds with similar factor loadings

WHAT ABOUT LOGIT?

- Koijen Yogo (1999), Koijen Richmond Yogo (2024):
“log utility + a specific factor model with only 1 factor equal to the current portfolio = logit”
- Logit accounts for substitution, so nothing missing
- **An arbitrary factor model is not equivalent to logit**
 - Logit: when the price of any bond increases, CalPERS replaces it proportionally to its existing portfolio
 - Factor model: when the price of a bond increases, CalPERS replaces it disproportionately with bonds with similar factor loadings

WHAT ABOUT LOGIT?

- Koijen Yogo (1999), Koijen Richmond Yogo (2024):
“log utility + a specific factor model with only 1 factor equal to the current portfolio = logit”
- Logit accounts for substitution, so nothing missing
- **An arbitrary factor model is not equivalent to logit**
 - Logit: when the price of any bond increases, CalPERS replaces it proportionally to its existing portfolio
 - Factor model: when the price of a bond increases, CalPERS replaces it disproportionately with bonds with similar factor loadings
- Both models satisfy our assumptions and hence can have relative elasticity estimated from the cross-section
 - Assuming logit-specific structure makes it enough to back out substitution
 - Analogy: if you assume no substitution at all, you would also get all \mathcal{E} from the cross-section

ESTIMATING SUBSTITUTION WITH THE TIME SERIES

WHY SUBSTITUTION MATTERS

Hengjie wants to know:

- Will CalPers maintain its green tilt if the price of green bonds become very expensive relative red bonds?

WHY SUBSTITUTION MATTERS

Hengjie wants to know:

- Will CalPers maintain its green tilt if the price of green bonds become very expensive relative red bonds?
- How much will CalPers size down its bond positions if all bond prices increase?

WHY SUBSTITUTION MATTERS

Hengjie wants to know:

- Will CalPers maintain its green tilt if the price of green bonds become very expensive relative red bonds?
- How much will CalPers size down its bond positions if all bond prices increase?
- Answer to these questions relies on knowing substitution!

SIMPLIFYING SUBSTITUTION

Under assumptions A1 and A2, can replace the asset-level problem of substitution with a portfolio-level problem:

- Re-package prices and demand:

$$\Delta P_{agg} = \frac{1}{N} \sum_i \Delta P_i,$$

$$\Delta P_X = \frac{1}{N} \sum_i X_i \Delta P_i$$

$$\Delta P_{idio,i} = \Delta P_i - \Delta P_{agg} - X_i \Delta P_X$$

$$\Delta D_{agg} = \frac{1}{N} \sum_i \Delta D_i$$

$$\Delta D_X = \frac{1}{N} \sum_i X_i \Delta D_i$$

$$\Delta D_{idio,i} = \Delta D_i - \Delta D_{agg} - X_i \Delta D_X$$

SIMPLIFYING SUBSTITUTION

Under assumptions A1 and A2, can replace the asset-level problem of substitution with a portfolio-level problem:

- Re-package prices and demand:

$$\Delta P_{agg} = \frac{1}{N} \sum_i \Delta P_i,$$

$$\Delta D_{agg} = \frac{1}{N} \sum_i \Delta D_i$$

$$\Delta P_X = \frac{1}{N} \sum_i X_i \Delta P_i$$

$$\Delta D_X = \frac{1}{N} \sum_i X_i \Delta D_i$$

$$\Delta P_{idio,i} = \Delta P_i - \Delta P_{agg} - X_i \Delta P_X$$

$$\Delta D_{idio,i} = \Delta D_i - \Delta D_{agg} - X_i \Delta D_X$$

- *Decompose the response of demand to prices into three univariate components:*

Relative:

$$\Delta D_{idio,i} = \hat{\mathcal{E}} \Delta P_{idio,i}$$

Meso:

$$\Delta D_X = \tilde{\mathcal{E}}_{agg} \Delta P_{agg} + \tilde{\mathcal{E}}_X \Delta P_X$$

Macro:

$$\Delta D_{agg} = \bar{\mathcal{E}}_{agg} \Delta P_{agg} + \bar{\mathcal{E}}_X \Delta P_X$$

AGGREGATION INTUITION

- Under assumptions A1 and A2:

$$\mathcal{E} = \hat{\mathcal{E}}\mathbf{I} + X\mathcal{E}_XX'$$

- $\hat{\mathcal{E}}$ is a scalar
 - \mathcal{E}_X is only $K \times K$ (factor model)
- Project the change in price along the factor X direction:

$$\begin{aligned}\Delta P_X &= (X'X)^{-1}X'\Delta P \\ &= \left(\hat{\mathcal{E}}(X'X)^{-1}X'\mathbf{I}_N + (X'X)^{-1}X'X\mathcal{E}_XX' \right) \Delta D \\ &= \left(\hat{\mathcal{E}}\mathbf{I}_K + \mathcal{E}_X(X'X) \right) \Delta D_X\end{aligned}$$

AGGREGATION INTUITION

- Two easy cases:
 - Only one characteristic ($K = 1$):

$$\check{\mathcal{E}} = \begin{pmatrix} \hat{\mathcal{E}} + N(\mathcal{E}_X)_{11} & N(\mathcal{E}_X)_{12} \\ N(\mathcal{E}_X)_{21} & \hat{\mathcal{E}} + N(\mathcal{E}_X)_{22} \end{pmatrix} = \begin{pmatrix} \bar{\mathcal{E}}_{agg} & \bar{\mathcal{E}}_X \\ \tilde{\mathcal{E}}_{agg} & \tilde{\mathcal{E}}_X \end{pmatrix}$$

- Observables are group dummies: consider the case when the observables are dummy variables for disjoint groups.

$$\Delta P_{X,k} = \frac{1}{N_k} \sum_{i \in k} \Delta P_i$$

ESTIMATING THE MESO AND MACRO ELASTICITIES

Meso:

$$\Delta D_X = \tilde{\epsilon}_{agg} \Delta P_{agg} + \tilde{\epsilon}_X \Delta P_X$$

Macro:

$$\Delta D_{agg} = \bar{\epsilon}_{agg} \Delta P_{agg} + \bar{\epsilon}_X \Delta P_X$$

- Substitution boils down to relation between aggregate and observable based portfolios
 - Response of overall demand and green portfolio tilt to aggregate bond price and price of green-minus-brown portfolio
 - Low dimensional
- Need joint instruments for prices **in time series**:
 - To estimate macro elasticity, need to account for simultaneous change in price of green-minus-brown
 - Only controlling for the price is generally a bad control (in particular if demand shocks are correlated)

ESTIMATING MULTIPLIERS: AN EMPIRICAL EXAMPLE

EXAMPLE: CORPORATE BOND RELATIVE MULTIPLIER

- U.S. investment-grade corporate bonds (following Chaudhary, Fu, Li, 2024)
- Steps to go through when conducting causal inference in asset pricing:
 - 1 choose a source of variation
 - 2 assess exogeneity
 - 3 assess assumptions A1 and A2 and select observables + units
 - 4 implement the regression analysis
- Step 1: flow-induced demand shock Z_{it} : fund flow in mutual funds \times portfolio composition (Coval Stafford 2007, Lou 2012)
- Step 2: assess exogeneity, i.e., $Z_{it} \perp \epsilon_{it} | X_{it}$
 - example threat to identification: another investor type (e.g., insurance companies) buying the same bonds held by mutual funds that receive a lot of inflows

STEP 3: DIAGNOSTIC FOR A1 – BALANCE ON COVARIANCES

Do treated bonds comove the same way with broad portfolios as the control bonds?

- 1 At each date t , form a long-short portfolio based on whether Z_{it} is above (“treated”) or below (“control”) the median
- 2 Compute the β of the long-short return on broad indices in a window around t (here: 2y)
- 3 β different from zero \Rightarrow substitution likely not homogeneous

STEP 3: DIAGNOSTIC FOR A1 – BALANCE ON COVARIANCES

Do treated bonds comove the same way with broad portfolios as the control bonds?

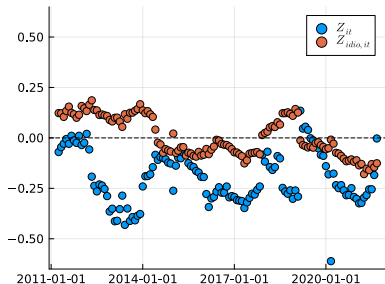
- 1 At each date t , form a long-short portfolio based on whether Z_{it} is above (“treated”) or below (“control”) the median
 - 2 Compute the β of the long-short return on broad indices in a window around t (here: 2y)
 - 3 β different from zero \Rightarrow substitution likely not homogeneous
-
- Treated and control bonds may differ systematically based on the observables, which may drive differences in β
 \rightarrow natural if investors choose their flows along dimensions like duration and credit risk
 - Do the treated and control comove the same way *conditional on observables*?
 - $Z_{idio,it}$: residual of instrument regressed on a date fixed effect, **duration** \times date fixed effects and **credit rating** \times date fixed effects

STEP 3: DIAGNOSTIC FOR A1 – BALANCE ON COVARIANCES

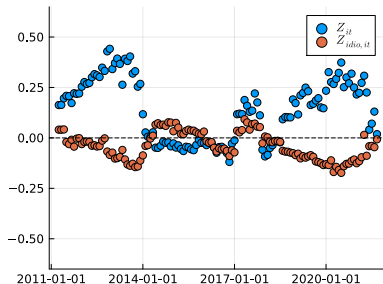
Do treated bonds comove the same way with broad portfolios as the control bonds?

- 1 At each date t , form a long-short portfolio based on whether Z_{it} is above (“treated”) or below (“control”) the median
 - 2 Compute the β of the long-short return on broad indices in a window around t (here: 2y)
 - 3 β different from zero \Rightarrow substitution likely not homogeneous
-
- Treated and control bonds may differ systematically based on the observables, which may drive differences in β
 \rightarrow natural if investors choose their flows along dimensions like duration and credit risk
 - Do the treated and control comove the same way *conditional on observables*?
 - $Z_{idio,it}$: residual of instrument regressed on a date fixed effect, **duration** \times date fixed effects and **credit rating** \times date fixed effects
 - Alternative unit to bond returns: yield changes ▶ A1 yield changes ▶ Multiplier yield changes
 - Similar diagnostic for A2: balance on idiosyncratic volatility ▶ A2 diagnostic

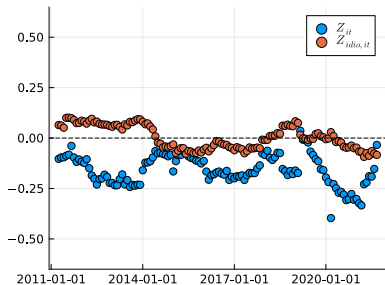
A. Corporate Bond Index



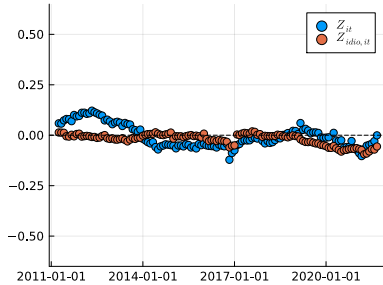
B. High—Low Credit Rating



C. Long—Short Term Bonds



D. Stock Index



STEP 4: IMPLEMENT THE REGRESSION

Relative multiplier $\widehat{\mathcal{M}} \approx 0$

	Return $\Delta P_{it}/P_{i,t-1}$				
	(1)	(2)	(3)	(4)	(5)
<i>Demand shock:</i>					
Z_{it}	1.541*	-0.254	0.019		
	(0.637)	(0.229)	(0.065)		
$Z_{idio,it}$				0.019	0.019
				(0.065)	(0.065)
Date Fixed Effects		Yes	Yes	Yes	Yes
Duration \times Date Fixed Effects			Yes	Yes	
Credit Rating \times Date Fixed Effects			Yes	Yes	
N	646,335	646,335	646,335	646,335	646,335
R^2	0.010	0.415	0.632	0.632	0.415

CONCLUSION

- I hope Hengjie is happy now

CONCLUSION

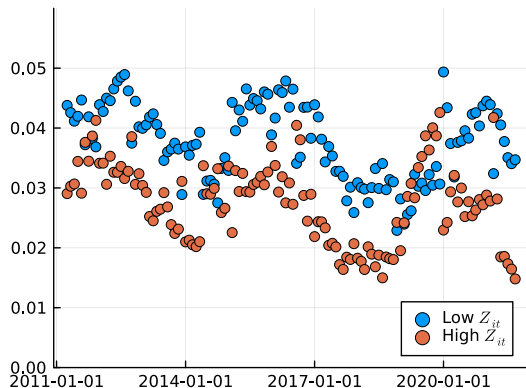
- Key challenge for causal inference in asset pricing: substitution across assets

CONCLUSION

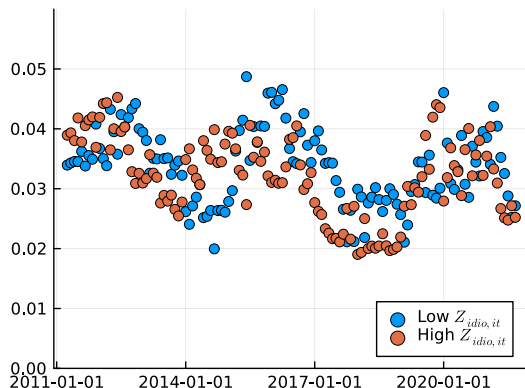
- Key challenge for causal inference in asset pricing: substitution across assets
- **An elementary condition for valid inference:** homogenous substitution conditional on observables
 - difference in substitution driven by a known set of observables
- **Standard cross-sectional causal inference** identifies relative elasticity or its inverse, relative multiplier
 - Guidance on designing settings such that assumptions are plausible
 - Compatible with usual covariance matrix assumptions
- **Time series identification with observable-based portfolios** reveals substitution
 - Need to consider all dimensions of substitution jointly

DIAGNOSTICS FOR A2 – BALANCE ON IDIOSYNCRATIC VOLATILITY

A. Idiosyncratic Volatility (Z_{it})

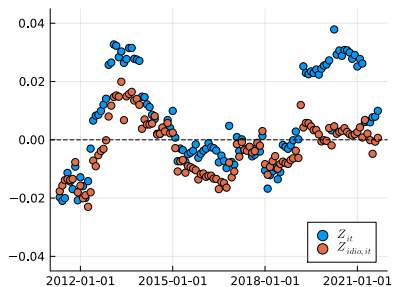


B. Idiosyncratic Volatility ($Z_{idio,it}$)

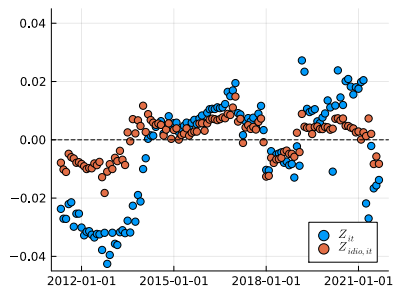


■ Average idiosyncratic volatility among treated versus control bonds

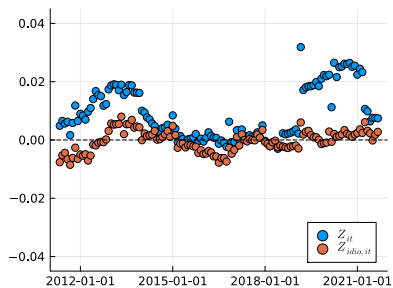
A. Corporate Bond Index



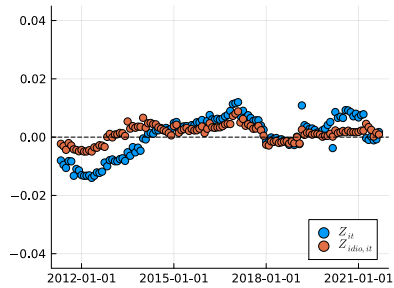
B. High—Low Credit Rating



C. Long—Short Term Bonds



D. Stock Index



Relative multiplier $\widehat{\mathcal{M}} = -0.072$

	Yield change ΔY_{it}				
	(1)	(2)	(3)	(4)	(5)
<i>Demand shock:</i>					
Z_{it}	-0.384*	-0.104*	-0.072**		
	(0.166)	(0.047)	(0.027)		
$Z_{idio,it}$				-0.072**	-0.072**
				(0.027)	(0.027)
Date Fixed Effects		Yes	Yes	Yes	Yes
Duration \times Date Fixed Effects			Yes	Yes	
Credit Rating \times Date Fixed Effects			Yes	Yes	
N	630,255	630,255	630,255	630,255	630,255
R^2	0.004	0.071	0.089	0.089	0.070