Collateral Demand in Wholesale Funding Markets

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Views are solely those of the authors and not the Bank of England.

Repo Markets: How they work

Repurchase agreements (repo):

- ▶ Borrower sells asset at t & promises to buy it back at t + 1.
- Collateralized lending.
- ► Lender temporarily owns asset.

Repo serves two functions:

- 1. Funding demand: Acquiring funding cheaply.
 - \rightarrow Collateral valued only as insurance.
- Collateral demand: Acquiring assets temporarily.
 - \rightarrow Usage of collateral valuable, eg to short.

Repo Markets: Why they matter

Important:

- lacktriangle Key wholesale funding market o financial stability.
- $lackbox{ Necessary input to a shorting trade }
 ightarrow$ asset prices.

Economic interest:

Organization of market with two functions.

Question

Does collateral function complement funding function?

- What happens to eq'm funding absent collateral demand?
- Does this effect vary over time or in crises?
- Implications for regulation and policy?

What we do

Our focus: distribution of collateral demand across firms.

- 1. Transaction data of repo against UK gov bonds with firm ids.
 - → Heterogeneity in repo rates across firms.
- 2. Equilibrium model of repo.
 - → Effect of heterogeneous collateral demand across firms.
- 3. Structurally estimate model.
 - \rightarrow Infer & interrogate firm-time-asset collateral demand.
 - → Counterfactual: remove collateral demand.

What we find

Does collateral function complement liquidity function?

No! Volumes and gains to trade higher absent collateral demand.

- ▶ Joint distribution of funding and collateral needs across firms.
- Firms that need funding are also those that value collateral.

Empirical literature on repo

Duffie (1996); Gorton and Metrick (2012); Copeland, Martin & Walker (2014); Krishnamurthy, Nagel & Orlov (2014); Mancini, Ranaldo & Wrampelmeyer (2016); Boissel, Derrien, Ors & Thesmar (2017); D'Amico, Fan & Kitsul (2018); Ranaldo, Schaffner & Tsatsaronis (2019); Hüser, Lepore & Veraart (2021); Eisenschmidt, Ma & Zhang (2022); Ballensiefen, Ranaldo & Winterberg (2023); Huber (2023).

Contribution

- 1. Structural measurement of collateral demand.
- 2. Distribution in XS and TS.
- 3. Equilibrium effects.
- 4. Negative effect on repo market functioning.

Empirical literature on repo: Specialness

Duffie (1996); Gorton and Metrick (2012); Copeland, Martin & Walker (2014); Krishnamurthy, Nagel & Orlov (2014); Mancini, Ranaldo & Wrampelmeyer (2016); Boissel, Derrien, Ors & Thesmar (2017); D'Amico, Fan & Kitsul (2018); Ranaldo, Schaffner & Tsatsaronis (2019); Hüser, Lepore & Veraart (2021); Eisenschmidt, Ma & Zhang (2022); Ballensiefen, Ranaldo & Winterberg (2023); Huber (2023).

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Empirical literature on repo: Structural estimation

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Empirical Facts

BoE transaction data on \approx universe of repo trading against UK government collateral (gilts) from 2017-23.

Facts on collateral demand:

- 1. Underlying asset matters for hedge funds, not MMFs.
- 2. Most repo rates below risk-free rate.
- 3. Hedge funds charge lower rates to lend.
- 4. Rates higher when collateral is interchangeable.

Background facts:

► Market power, exogenous networks, interdealer trade, etc.

Rate Variation: Hedge Fund vs MMF Lending

Table reports R^2 in regression of reporates on FE for firm type.

Fixed effects	Hedge fund	MMF	
Week-Maturity	0.50	0.31	
Week-Maturity-Borrower	0.56	0.98	
Week-Maturity-Lender	0.62	0.42	
Week-Maturity-Asset	0.94	0.73	

What about:

- 1. q?
- 2. confounding factors?
- 3. quantification?
- 4. counterfactuals?
- ightarrow model

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Model: Setup

Assets & Agents

- $ightharpoonup \mathcal{A}$ assets, indexed by a: exchange cash for collateral.
- ▶ Return to funding for agent $i \sim N(\nu_i, 1)$.
- ▶ Return to collateral for agent $i \sim N(\eta_i^a, \sigma)$.
- ▶ Mean-var preferences with risk aversion κ .

Trading structure

- $ightharpoonup N_d$ dealers and N_c customers on fixed network \mathbf{G}^a .
- Firm k has set \mathcal{N}_k^a as neighbours.
- No customer-customer links.
- 1. Competitive interdealer market indexed by *D*.
- 2. Dealer-customer trade, where dealers have market power.

Model: Setup

Trading

- $ightharpoonup q_{ii}^a$ borrowing by *i* from *j* against *a*.
- ▶ $Q_i^a = \sum_{j \in \mathcal{N}_i^a} q_{ij}^a$ total net borrowing by *i* against *a*.
- ▶ $Q_i = \sum_a Q_i^a$ total net borrowing by *i*.
- $ightharpoonup r_{ii}^a$ interest rate.
- $ightharpoonup \epsilon_{im}^a$ non-pecuniary, relationship-specific benefits.

Payoff to firm i

$$\underbrace{\nu_i Q_i - \frac{\kappa}{2} Q_i^2}_{\text{Funding}} \underbrace{-\sum_{a} \eta_i^a Q_i^a - \sum_{a} \frac{\kappa}{2} \sigma(Q_i^a)^2}_{\text{Collateral demand}} - \underbrace{\sum_{a} \sum_{m \in \mathcal{N}_i^a} q_{im}^a(r_{im}^a + \epsilon_{im}^a)}_{\text{Transaction terms}}$$

First order condition

Customer j, with respect to quantity q_{ij}^a :

$$-\nu_{j} + \kappa Q_{j} + \eta_{j}^{a} + \kappa \sigma Q_{j}^{a} + r_{ij}^{a} = 0$$
-j's MB from cash j's MB from collateral

Dealer i, with respect to quantity q_{ij}^a :

$$\nu_i - \kappa Q_i \qquad - \left(\eta_i^{\it a} + \kappa \sigma Q_i^{\it a}\right) \qquad - \kappa \sum_{\it l} q_{ij}^{\it l} - \kappa \sigma q_{ij}^{\it a} - \epsilon_{ij}^{\it a} - r_{ij}^{\it a} = 0$$
 i's MB from cash -i's MB from collateral Price effect

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Equilibrium

Solution:

Linear FOCs where network link exists, given **G**.

Equilibrium quantity q_{ij}^a depends on:

- ▶ Relative counterparty characteristics: v_i, v_j and η_i^a, η_i^a .
- Network: counterparties' counterparties' characteristics, etc.

Effect of collateral demand on gains to trade $(\eta_i^a = 0, \forall i)$:

- Correlation between funding and collateral demand across i.
- ► Therefore an empirical question. Example

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Estimation: Setting

Task is to recover as flexibly as possible

- funding demand ν_{it} ;
- ightharpoonup collateral demand η_{it}^a ;
- ightharpoonup risk σ ; and
- ightharpoonup risk aversion κ ;

from

- ightharpoonup observed quantities q_{iit}^a ; and
- ightharpoonup observed rates r_{ijt}^a .

Estimation: Overview

Model: Dealer i FOC with respect to q_{ijt}^a :

$$r_{ijt}^{a} = \underbrace{\nu_{it} - \kappa Q_{it}}_{i'\text{s MB from cash } -i'\text{s MB from collateral}} \underbrace{-\kappa \sum_{l} q_{ijt}^{l} - \kappa \sigma q_{ijt}^{a}}_{\text{Price effect}} - \epsilon_{ijt}^{a}$$

Two step estimation:

- 1. Infer (κ, σ) from variation across j, within i t.
- 2. Given these estimates, infer (ν_{it}, η_{it}^a) from variation across a.

Challenges:

- Simultaneity: Gilt prices and trading patterns by firm as IV.
- ▶ Level identification: $\eta_{it}^a = 0$ when a is "general collateral".



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Results

Variation across firms:

- 1. Variation across type: banks and HF have high η .
- 2. Positive correlation across firms between η and ν .

Variation across time:

- 3. Funding demand tracks central bank rate.
- 4. Level and dispersion in collateral demand track volatility.

Implication:

Collateral demand bad for funding, particularly in stress?



Variation in Funding & Collateral Demand

Most variation across firms, not across assets:

Fixed Effects	Funding demand	Collateral demand	
Time t	0.96	0.07	
Firm i	0.14	0.49	
Asset a		0.05	
Firm-Asset ia		0.58	
Firm-Time it		0.85	
Asset-Time at		0.19	

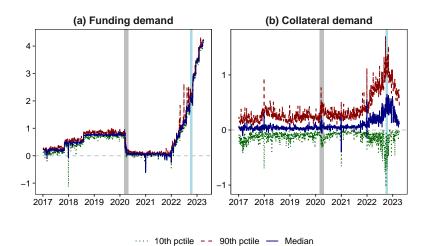
Variation across firm types

	Funding demand $ u_{it}$ (1)	Collateral demand η_{it}^a (2)
Bank	0.68***	0.13***
	(0.007)	(0.0007)
Dealer	0.81***	0.23***
	(0.006)	(0.0004)
Fund	0.84***	0.07***
	(0.005)	(0.001)
Hedge Fund	0.70***	0.11***
	(0.004)	(0.0007)
MMF	0.61***	0.05***
	(0.01)	(0.003)
Other	0.77***	0.13***
	(800.0)	(0.002)
PFLDI	0.71***	-0.08***
	(0.006)	(0.001)
R^2	0.005	0.05
Observations	167,037	1,490,509

Correlation between funding and collateral demand

	Collateral demand η_{it}^a		
	(1)	(2)	(3)
Funding demand $ u_{it}$	0.20*** (0.0003)	0.95*** (0.001)	0.12*** (0.02)
R ² Observations	0.22 1,563,051	0.74 1,563,051	0.57 1,563,051
Day FEs Firm FEs		Yes	Yes

Variation over time



Results

Variation across firms:

- 1. Variation across type: banks and HF have high η .
- 2. Positive correlation across firms between η and ν .

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Counterfactual: Removing Collateral Demand

Removing collateral demand:

- ▶ Set $\eta_{it}^a = 0$ for all a, i, t.
- Collateral equally useful for everyone only as insurance.

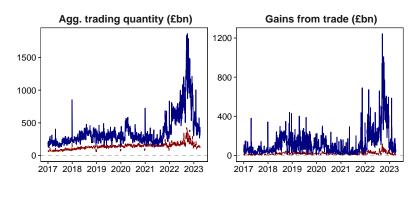
Effect, relative to baseline:

Volumes and gains to trade higher, particularly in stress.

Extension, wrt correlation:

- Rearrange η_{it}^a across *i* to reverse correlation.
- ▶ Undertake same counterfactual removing collateral demand.
- ▶ Effect reversed: this is about correlation.

Counterfactual: Quantities & GTT



- - Collateral demand — No collateral demand

Role of Correlation

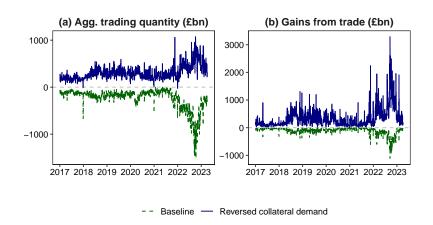


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Regulation

Problem: banks cannot simultaneously manage risk and funding.

- ▶ Banks need to be long on bonds to fund themselves...
- ... when they want to reduce inventory risk.

Implications for regulation/policy?

- Uncovered short-selling.
- Central bank repo accepting other collateral.
- Central bank collateral swap facilities.
- Monetary policy.

Conclusion

- Collateral demand is a key driver of repo outcomes.
- Effect depends on joint distribution with funding demand.
- Finding: dual repo functions do not always combine well.

Conclusion

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- Effect depends on joint distribution with funding demand.
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Thank you! patrick.coen@tse-fr.eu

Annexes

Background facts

Trade details:

- Mostly short maturity.
- Fully or over collateralized, no default.

Trade structure:

- ► Network sparse & broadly fixed. Details
- ► Dealers earn a spread. Dealer spreads
- D-D trade mostly on platforms, D-C trade OTC.

Firm types:

- MMFs uniquely lend, do not use collateral. Details
- Hedge funds borrow & lend, and may use to short.
- Different firms borrow against different gilts. Wallet variation

Net lending by sector

	Trade Share (%)	Daily net lending (%)	Daily net lending (£bn)
Dealer	66.1	-3.8	-4.6
Bank	11.7	-31.4	-7.5
Hedge Fund	10.3	-0.2	-0.4
Fund	4.2	62.5	5.2
MMF	2.9	97.4	6.2
PFLDI	2.8	18.9	0.9
Other	2.0	0.6	0.5

Additional facts

- 1. Fewer than 2% of counterparty pairs have non-zero trade in the whole sample.
- 2. Over 95% of transactions after January 2022 onwards were between traders who had traded together before January 2022.

 Back

Repo rate variation

Fixed effects	R-squared
Deal characteristics	
Week	0.37
Week-Asset	0.86
Week-Maturity	0.42
Week-Asset-Maturity	0.90
Trader characteristics	
Week-Borrower	0.51
Week-Lender	0.45
Week-Borrower-Lender	0.59

Rate variation

Dealer spreads

	(1)	Repo rate (% (2)) (3)
Dealer lending	0.155***	0.149***	0.092***
	(0.007)	(0.002)	(0.0006)
R^2 Observations	0.23	0.35	0.81
	1,003,270	1,003,270	1,003,270
Week FEs Week-Dealer FEs Week-Dealer-Asset FEs	Yes	Yes	Yes

Repo Rates & Collateralization Type

	Repo rate (%)			
	(1)	(2)	(3)	(4)
General Collateral	0.09***	0.09***	0.09***	0.10***
	(0.006)	(0.01)	(0.003)	(0.004)
R^2	0.30	0.20	0.55	0.43
Observations	6,095,617	6,095,617	6,095,617	6,095,617
Week FEs	Yes			
Borrower-Lender FEs		Yes		
Borrower-Week FEs			Yes	
Lender-Week FEs				Yes



Rates for hedge funds vs MMFs

	Repo rate (%)			
	(1)	(2)	(3)	(4)
Lender: Hedge fund	-0.06***	-0.08***	-0.003***	-0.002**
	(0.006)	(0.003)	(0.001)	(0.001)
R^2	0.38	0.58	0.94	0.97
Observations	371,649	371,649	371,649	371,649
Week FEs	Yes			
Borrower-Week FEs		Yes		
Borrower-Asset-Week FEs			Yes	
Asset-Mat-Borr-Week FEs				Yes



Model: Simplified example

One dealer i, one customer j, one asset:

- Equilibrium net borrowing by i:

$$q_{ij} = \frac{\Delta \nu - \Delta \eta}{3\kappa (1+\sigma)}$$

Equilibrium trading volume:

$$|q_{ij}| = \frac{|\Delta \nu - \Delta \eta|}{3\kappa(1+\sigma)}$$

Gains to trade:

$$GTT = \frac{2(\Delta \nu - \Delta \eta)^2}{9\kappa(1+\sigma)}$$

Model: Simplified example

One dealer i, one customer j, one asset:

- - $\rho \in [-1 \ 1]$: correlation btw liquidity and collateral demand.
 - ▶ $\bar{\eta} \in [0 \ 1]$: magnitude of collateral demand.
- **Effect** of collateral demand on GTT depends on correlation ρ :

$$\frac{dGTT}{d\bar{\eta}} \quad \begin{cases} >0, & \text{if } \rho < 0 \\ <0, & \text{otherwise} \end{cases}$$

► Effect of collateral demand therefore an empirical question.

Estimation: Step 1

Estimating equation:

$$r_{ijt}^{a} = \delta_{it}^{a} - \left[\kappa \sum_{l} q_{ijt}^{l} + \kappa \sigma q_{ijt}^{a}\right] \mathbb{1}_{ij} + \epsilon_{ijt}^{a}$$

where $\mathbb{1}_{ij} = 1$ if i has market power wrt j.

Identification:

- ightharpoonup Challenge: standard joint determination of q and r.
- ▶ Different j trade different a (exogenous "wallet").
- ▶ Change in price of gilt a exogenous to ϵ_{iit}^a .
- ► Shift-share IV: lag wallet shares, interact with price.

Estimation Details

Estimation: Step 2

Model:

$$\delta_{it}^{a} = \nu_{it} - \kappa Q_{it} - \eta_{it}^{a} - \kappa \sigma \sum_{m} q_{imt}^{a}$$

Second step estimation:

$$\hat{\delta}_{it}^{a} + \hat{\kappa}\hat{\sigma}\sum_{m}q_{imt}^{a} + \hat{\kappa}Q_{it} = \nu_{it} - \eta_{it}^{a}$$

- Decompose network-adjusted average interest rates for i.
- Level identification from following assumption:

$$\eta_{it}^{GC} = 0 \quad \forall i, t$$

Instruments: Details

Instruments:

$$egin{aligned} z_{1,jt} &= \sum_{a \in \omega_j} s^a_{jt} imes ext{price}^a_t \ z^a_{2,jt} &= z_{1,jt} - s^a_{jt} imes ext{price}^a_t \end{aligned}$$

First stage:

$$q_{ijt}^{a} = \alpha_{it}^{a} + \beta_{1}z_{1,jt} + \beta_{2}z_{2,jt}^{a} + e_{ijt}^{a}$$

$$\sum_{l} q_{ijt}^{l} = \alpha_{it}^{a} + \beta_{3}z_{1,jt} + \beta_{4}z_{2,jt}^{a} + e_{ijt}^{a}$$

Second stage:

$$r_{ijt}^{a} = \delta_{it}^{a} - \left[\kappa \sum_{l} q_{ijt}^{l} + \kappa \sigma q_{ijt}^{a}\right] \mathbb{1}_{ij} + \epsilon_{ijt}^{a}$$

Back

Estimates: risk & risk aversion

	Repo rate r_{ijt}^a (%)		
	OLS	2SLS	
	(1)	(2)	
$\sum_{l} q_{ijt}^{l}$	-0.01***	-0.02***	
	(0.0009)	(0.002)	
q_{ijt}^a	-0.12***	-0.18***	
9 -	(0.002)	(0.003)	
Wald (1st stage), $\sum_{l} q_{iit}^{l}$		6,377.2	
Wald (1st stage), $\sum_{l} q_{ijt}^{l}$ Wald (1st stage), q_{ijt}^{a}		2,170.8	
R^2	0.996	0.997	
Within R ²	0.027	0.037	
Observations	599,384	527,295	
Firm-asset-day FEs	Yes	Yes	
Firm-counterparty FEs	Yes	Yes	



First Stage

	q_{ijt}^a OLS (1)	$\frac{\sum_{l} q_{ijt}^{l}}{2SLS}$ (2)
$z_{1,jt}$ $z_{2,jt}^a$	-0.0114*** (0.0002) 0.0116*** (0.0002)	-0.0072*** (0.0002) 0.0009*** (0.0002)
R ²	0.80069	0.86838
F-test	535.18	878.98
Observations	527,295	527,295
Firm-asset-week FEs	Yes	Yes
Firm-counterparty FEs	Yes	Yes



Collateral Demand & Asset Prices

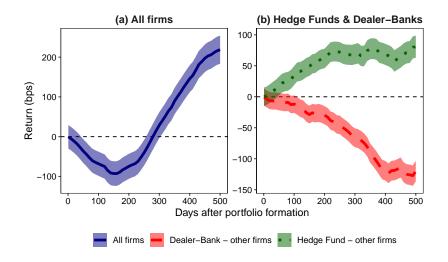
Questions:

- Why do banks have collateral demand?
- Does collateral demand predict future bond prices?
- Is collateral demand about hedging or speculation?

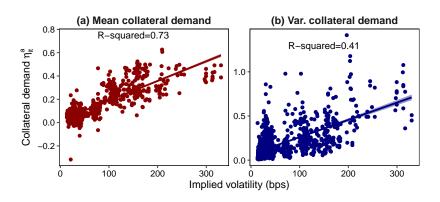
Approach:

▶ Go short (long) on bonds with high (low) collateral demand.

3. Collateral Demand & Asset Prices



Volatility & Collateral Demand



Back

Sector heterogeneity

	Trade Share (%)	Daily net lending (%)	Daily net lending (£bn)
Dealer	66.1	-3.8	-4.6
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Week-Borrower-Lender	0.59

Rate variation by firm type

Fixed effects	Hedge fund	MMF	
Week-Maturity	0.50	0.31	
Week-Maturity-Borrower	0.56	0.98	
Week-Maturity-Lender	0.62	0.42	
Week-Maturity-Asset	0.94	0.73	

Rates for general collateral

	Repo rate (%)			
	(1)	(2)	(3)	(4)
General Collateral	0.09***	0.09***	0.09***	0.10***
	(0.006)	(0.01)	(0.003)	(0.004)
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Borrower-Week FEs			Yes	
Lender-Week FEs				Yes

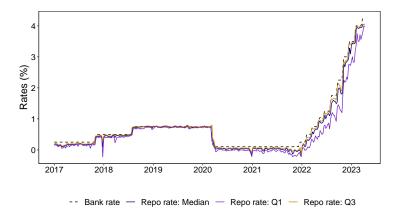
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Week FEs Week-Dealer FEs Week-Dealer-Asset FEs	Yes	Yes	Yes

Rates through time on dealer repo lending



Regression Results

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Regression Results: First Stage

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Implied volatility

