



# What Drives Global Corporate Bond Returns?

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# Outline

- Intro
- Main Finding
- Comments
  - Economic meaningful: PnL, Transaction Cost
  - Local vs Global factor comparison



# Intro

## Academic Background

- PhD in Economics at Princeton (2022)
- Thesis: “Essays on Institutional Ownership and Liquidity in the Corporate Bond Market”
  - "Investor Composition and the Liquidity Component in the U.S. Corporate Bond Market" - Forthcoming in Journal of Finance
  - "Investor Concentration, Liquidity and Bond Price Dynamics" - Working Paper

## Industry Experience

- Lead Applied Scientist, building ML models at Snowflake
- Global Macro Quant Researcher at Two Sigma (2022–2025)
- Quant Research Intern, Corporate Bond/ETF desk at Susquehanna International Group (2020–2021)



# Main Finding

Global Corporate Bond Return Drivers:

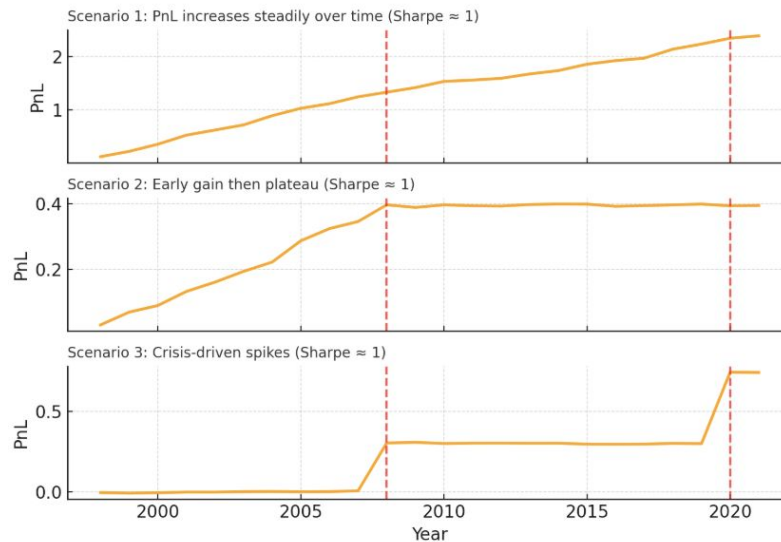
- Signal Research in Corporate Bonds (extend to Global Markets):
  - Bond and firm-level signals
  - Effective signals were primarily equity-related: Past equity return, return around earnings, DEsprd
    - Buy bonds with higher EQ return
- Global market factor vs Other factors models
  - Global market factor performs better than other multiple factors models
  - Implication: Useful for hedging

# Comments - Economically Meaningful

- What does the P&L look like?

return spreads and alphas at the 5% level. In particular, EQret.1.0 demonstrates predictive power at the 1% significance level, which cannot be accounted for by commonly used factor models, and the associated Sharpe ratio is approximately 1.5 for both unhedged and hedged returns. Additionally, a fourth characteristic related to past stock returns, EQret.ind.6.1, also generates significant alphas

PnL Scenarios (1998-2021) — Adjusted for Similar Sharpe Ratios (~1)



# Comments - Economically Meaningful

- What is the turnover of the portfolio
  - Corporate bonds are very expensive to trade
  - 30 to 50 bps bid-ask spread for US IG bond (source: WRDS)
- What does the P&L look like after TC?
  - 6m EQ return may perform better than 1m EQ return
- Need a much higher Sharpe ratio for meaningful net returns

Median Bid-Ask Spread (%) by Credit Rating

(Bonds with Amount Outstanding > \$100 Million)

Rating	Bid-Ask Spread (%)	
AAA	0.35	
AA	0.29	
A	0.33	
BBB	0.38	
BB	0.42	
B	0.42	
CCC	0.62	
CC	1.28	
C	1.68	
D	1.70	↓



# Comments - Economically Meaningful

- Use a “liquid” and representative investment universe:
  - Example: US/EU IG bonds with a minimum outstanding amount of \$100 million
  - One representative bond per firm
- Data quality and reliability:
  - Uses actual TRACE transaction data over indicative quotes.
  - Data filtered (e.g., Kalman Filter, EMA)
- Keep bonds that are default -> survival bias

We apply a series of filtering criteria to obtain qualified bond-month observations. Specifically, we exclude bonds or observations that: (i) have a quoted price below 5 or above 1,000 for a face value of 100; (ii) have a remaining maturity of less than one year; (iii) are priced via matrix pricing; (iv) are **in default**; (v) feature floating coupon rates or are perpetual; (vi) are puttable or convertible; (vii) are structured notes, mortgage-backed, asset-backed, agency-backed, or equity-linked instruments; (viii) are secured or junior/subordinated; and (ix) are issued by government entities or financial institutions. We first apply criteria (i) through (iv) to daily quotes and then

# Comments - Global vs Local Factor

“Global factor explains more of the portfolio returns than local + foreign factors.”

- Surprising, given the segmentation and home bias of local bond markets
- Local bonds are less liquid and have more measurement error.
  - In the EU (the second most liquid credit market), local + foreign is actually better than Global.
- A stricter filter for the investment universe (keeping the most liquid bonds) may revise the table's findings.
- HYG, LQD or CDX Index return as factor
  - Better data quality
  - Widely used for hedging aggregate credit market risk

Table VII—Continued

Panel B: Performance of global, local, and local+foreign versions of factor models

		MKT		FF3		GB3		5-factor	
		Sig.	$\alpha$	Sig.	$\alpha$	Sig.	$\alpha$	Sig.	$\alpha$
US	G	10	0.16	14	0.17	10	0.15	9	0.13
	L	10	0.18	10	0.14	15	0.22	11	0.13
	L+F	10	0.18	11	0.14	16	0.23	9	0.13
EU	G	6	0.10	6	0.10	6	0.10	5	0.09
	L	9	0.12	8	0.09	6	0.10	9	0.09
	L+F	4	0.09	8	0.09	4	0.08	6	0.09





## Other Comments

- Compare bond signals and company signals within the same investment universe
- Account for implementation lag and one-day data:
  - Avoid using the day  $t$ 's closing price as an input signal to predict the close-to-close return between  $t$  to  $t+1$
  - For US credit, liquidity drops sharply after 3:30 pm ET to 4:00 pm ET
- Asymmetric effects between equity and bond markets
  - Bond analysts tend to be more pessimistic due to a focus on downside risk, contrasting with the general optimism of stock analysts
  - Would like to see the analysis of asymmetric impact of positive/negative returns between equity and credit

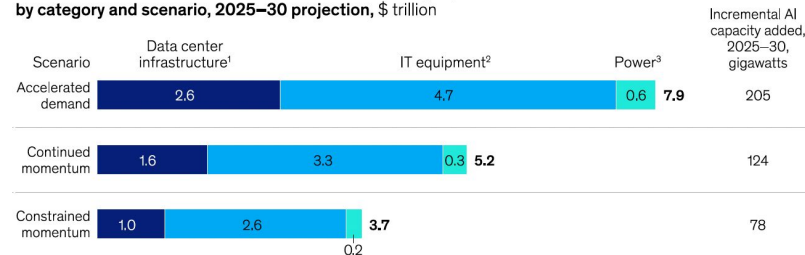
# Exciting Problems for FI - AI CapEx

- Post-2024: massive corporate issuance to fund AI CapEx (data center, power, GPU)
- Corporate bond markets + ABS/CMBS may finance a significant portion of **\$5–7T in AI-related infrastructure** (data centers, power, semiconductors) in 2025-2030 (source GS, McKinsey)
- This raises new fixed-income questions:
  - How to price tech infrastructure risk (low collateral, high CapEx)?
  - How will rating agencies rate OpenAI-related entities their investment in data center?
  - Could this create a new “AI bond bubble”?
- A large supply shock - who will buy them?

Exhibit 2

**Capital investments to support AI-related data center capacity demand could range from about \$3 trillion to \$8 trillion by 2030.**

**Global data center total capital expenditures driven by AI, by category and scenario, 2025–30 projection, \$ trillion**



Note: Figures may not sum to totals, because of rounding.

<sup>1</sup>Excludes IT services and software (eg, operating system, data center infrastructure management), since they require relatively low capex compared with other components.

<sup>2</sup>Includes server, storage, and network infrastructure. IT capex also accounts for replacing AI accelerators every 4 years.

<sup>3</sup>Assumes \$2.2 billion–\$3.2 billion/gigawatt (including power generation and transmission cost) to account for a range of power generation scenarios (eg, fully powered by gas, a combination of gas power and storage, and solar) and regional cost differences. Distribution cost is neglected, as most AI centers are expected to be >50 megawatt scale and connected to a transmission grid.

Source: McKinsey Data Center Capex TAM Model; McKinsey Data Center Demand Model



## Exciting Problems for FI - Generative AI's Impact on Corporate Bond Investing

- Productivity Gains:
  - Easier information processing and research, especially for FI
  - Improved risk management; tools (agents) will eventually emerge to assist in data collection and analysis
- Enhanced Signal Generation:
  - Access to more alternative data and signals from both structured and unstructured sources
  - Short-run: More profits for hedge funds and asset managers with access to the models and data
  - Long-run: Signal will decay
- Market Implications:
  - What are the consequences for market segmentation?
  - Ultimately, market efficiency will increase



# Summary

- Great paper and great pleasure to read it
  - Better and more accurate measurement of return and price, focus on a more liquid investment universe
  - Show PnL after TC
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