
”Leverage-based Asset Pricing”

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Leverage-based Asset Pricing

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Context

Leverage-based asset pricing: extract information about asset prices from the debt side of firms' balance sheets

- information can be encoded in firm characteristics such leverage ratios, debt structures, corporate bond prices, loan prices etc.

Application

Apply 'Leverage-based asset pricing' to the valuation of **private firms**

- investments in portfolios of private firms run through private equity firms are an increasingly important asset class for institutional investors

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Q: Can we learn about the valuation of private equity from prices in debt markets?

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Use modern asset pricing tools to extract information about the valuation of **private firms** from debt market information

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Q: Can we use '**credit market equivalent**' (CME) valuation to value private firms?

Credit Market Equivalents and the Valuation of Private Firms

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- it tells us the value of a payoff to an investor in a particular state in the future (here at horizon j)
 - if times are good at time $t + j$, payoffs are less valuable (\mathcal{M}_{t+j} lower)
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- an asset pricing model is a specification of an SDF
 - under some assumptions it exists and is unique (for example, no arbitrage)

Asset Pricing in Theory 101 and 1/2

If the payoff $X_{t+j} = D_{t+j} + P_{t+j}$, an SDF gives

$$P_t = E_t[\mathcal{M}_{t+j}D_{t+j} + P_{t+j}]$$

and so, for any return $R_{t+j} = \frac{D_{t+j} + P_{t+j}}{P_t}$

$$1 = E_t[\mathcal{M}_{t+j}R_{t+j}]$$

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- if all assets satisfy the equation, they are priced consistently from investors' perspective
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Applications:

- $R \rightarrow \mathcal{M}$
if we observe returns, we can figure out what \mathcal{M} makes the equation hold
- $\mathcal{M} \rightarrow R$
if we have an \mathcal{M} we can check whether returns satisfy the equation

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 - In PE, we do not observe returns, as there is little trading
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$$\mathcal{M}_{t+1} = \exp(-R_{m,t+1})$$

- this public market equivalent (PME) valuation is based on the SDF of an investor who is fully invested in the public stock market
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- Korteweg and Nagel (2016) propose to value PE investments through

$$\mathcal{M}_{t+1} = \exp(a - bR_{m,t+1})$$

- this generalized public market equivalent (GPME) valuation is based on the SDF of an investor who is invested in the public stock market *and a risk free asset*
- Q: would investing in PE improve the investor's risk-return relation?

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In practice, we replace returns on the stock market portfolio $R_{m,t+1}$ with returns on a number j of credit market instruments, $R_{cj,t+1}$

Our stochastic discount factor therefore takes the form

$$\mathcal{M}_{t+1} = \exp(a - \sum_j b_j R_{cj,t+1})$$

Q: would investing in PE improve the risk-return relation of an investor **already invested in private firms' debt instruments**?

This question makes sense if equity and debt markets are not entirely segmented

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- We build portfolios of debt returns sorted on portfolio company characteristics and ask what factors $R_{cj,t+1}$ help best explain the cross-section of debt returns on PE portfolio companies in our sample
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 - we use LASSO variable selection techniques to identify our factors $R_{cj,t+1}$
- This standard asset pricing procedure yields our SDF

$$\mathcal{M}_{CME,t+1} = \exp\left(a - \sum_j b_j R_{cj,t+1}\right)$$

- we use it to compute CME valuations of PE firms' cash flows and compare them to PME and GPME

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- Credit-market equivalents of momentum, volatility, value, and liquidity price the cross-section of PE debt market returns
- CME valuation of private equity funds paints a more positive picture of PE performance than PME/GPME valuation in our sample
 - significant underperformance under PME/GPME turns positive or insignificant under CME
 - suggestive of significant exposure of buyout performance to credit factors unrelated to public stock market activity

Data

- PE data: cash flow and valuation data from 121 funds with about 2,400 unique sample portfolio companies with vintage years 1996-2013
 - some portfolio companies exhibit unrealized returns and latest valuations are self-reported
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- Loan data: from LPC, monthly data for 22,032 facilities from 6,610 borrowers, between 1998 and 2018
 - provides bid and ask quotes averaged across all dealers, type of facility (term versus revolver), merged with DealScan providing original amount, currency, country of origination, seniority condition, loan secured, loan base rate, and spread
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 - add information from ORBIS on private firm accounting data
- Merged data: Loan pricing data for about eighty percent of the funds
 - funds in merged are slightly smaller than benchmark Preqin sample
 - caveats:..... many!

Result #1: Debt Valuations predict PE performance

| | All PE deals | | |
|--|---------------------|---------------------|--------------------|
| | Exit uc | log(VM) | Holding R_E |
| | (1) | (2) | (3) |
| Distress loan | 0.514*** (0.078) | | |
| Log average bid price | | 3.824*** (0.614) | |
| Holding loan return (in %) (holding time as equity) | | | 1.182* (0.632) |
| Loan maturity (yrs) | -0.004 (0.013) | -0.155* (0.077) | 11.019* (5.431) |
| # of quotes | -0.021 (0.020) | 0.262*** (0.090) | 11.656 (15.875) |
| Fund size (m) | 0.000 (0.000) | -0.000 (0.000) | 0.003 (0.003) |
| Equity investment year FE | Yes | Yes | Yes |
| Industry FE | Yes | Yes | Yes |
| Observations | 299 | 299 | 82 |
| Adj. R^2 | 0.272 | 0.231 | 0.213 |

Estimates of linear probability model of **exit equity performance**

- debt performance predicts predict exit equity performance

Result #1: Debt Valuations predict PE performance

| | (1) | (2) | (3) with distressed loan |
|------------------------------|---------------------|---------------------|-----------------------------------|
| Distress loan | 1.335* (0.229) | | |
| Loan maturity (yrs) (t-1) | 0.451*** (0.044) | 0.451*** (0.043) | 0.556*** (0.106) |
| # of quotes (t-1) | 0.922** (0.036) | 0.923** (0.037) | 0.936 (0.081) |
| Log average bid price (t-1) | | 0.987*** (0.003) | 0.980*** (0.005) |
| Number of fund-deal-quarters | 5,856 | 5,856 | 1,753 |

Estimates of **hazard ratios** associated with GPs' decision to sell/hold portfolio investments

- debt performance predicts exit hazard rates

Result #1: Debt Valuations predict PE performance

| | (1) | (2) | (3) | (4) |
|---------------------|---------------------|---------------------|----------------------|----------------------|
| Taken private | -7.110** (2.913) | -6.007** (2.569) | -13.601* (7.014) | -9.853 (5.911) |
| Leverage | | -3.936* (2.301) | | -5.251* (3.109) |
| Growth | (2.913) | (2.569) | (7.014) | (5.911) |
| Stock return | | | 2.389*** (0.810) | 2.256*** (0.739) |
| Loan maturity (yrs) | | | 10.545*** (2.511) | 10.023*** (2.140) |
| # of quotes | | | -0.910* (0.465) | -0.829* (0.467) |
| Fund size (m) | | | 1.347 (1.219) | 1.128 (1.155) |
| | | | -0.000 (0.000) | -0.000 (0.000) |
| Inv Qtr-Year FE | Yes | Yes | Yes | Yes |
| Qtr-Year FE | Yes | Yes | Yes | Yes |
| Industry FE | Yes | Yes | Yes | Yes |
| Adj. R ² | 0.352 | 0.362 | 0.102 | 0.122 |
| Observations | 1,020 | 1,020 | 728 | 728 |

Regression estimates of avg bid loan prices one quarter after the company is taken private by a private equity firm

- being taken private significantly predicts loan prices *negatively*

Result #2: Credit Factors price Debt Returns

What determines loans returns? We sort loans on PE firms into quintile portfolios according to loan characteristics and compute portfolio excess returns

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We consider the following characteristics

- spread-to-maturity ('credit spread')
- momentum
- price as a percentage par value ('market-to-book')
- market capitalization ('size')
- volatility
- # of quotes
- bid-ask spreads ('liquidity')
- accruals
- profitability

We find significant return spreads (around 1 to 2 percent annually) associated with these characteristics

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We use high-low quintile portfolios (Q5-Q1) sorted on characteristics as candidate factors

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We then follow the standard two-stage regression approach

- run time series regressions of portfolio returns on factors to get exposure (β)
- check whether average portfolio returns line up with exposures in a cross-sectional regression to get risk price (λ)
- $E[R_{c,t+1} - r_f] = \sum_j \beta_{c,j} \lambda_{c,j}$?

We use to LASSO variable selection techniques to identify the relevant factors

Result #2: Credit Factors price Debt Returns

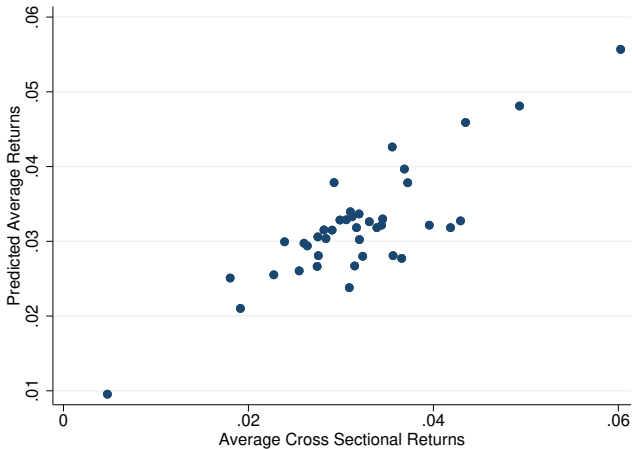
| | (1) rmrf β / SE |
|-------------|-----------------------------|
| Q5mQ1_mom | 0.018*** (0.004) |
| Q5mQ1_vola | 0.017*** (0.004) |
| Q5mQ1_price | -0.014*** (0.003) |
| Q5mQ1_MV | -0.014*** (0.003) |
| Q5mQ1_BA | 0.014*** (0.003) |

| | |
|------------|-------|
| Adj. R^2 | 0.596 |
|------------|-------|

Results from the second stage regression suggest a five-factor model for debt returns

- momentum, volatility, price ('market-to-book'), market cap ('size'), bid-ask ('liquidity') are significant debt return predictors

Result #2: Credit Factors price Debt Returns



The five-factor model prices quintile portfolios sorted on characteristics well

Result #3: Valuing PE

What about the \mathcal{M} ?

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Our SDF becomes

$$\mathcal{M}_{t+1} = \exp\left(a - \sum_j b_j R_{c_j,t+1}\right)$$

with

$$R_{c,t+1} = \begin{bmatrix} \text{Q5mQ1mom}_{t+1} \\ \text{Q5mQ1vola}_{t+1} \\ \text{Q5mQ1price}_{t+1} \\ \text{Q5mQ1MV}_{t+1} \\ \text{Q5mQ1BA}_{t+1} \end{bmatrix}$$

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We can use this SDF to price PE cash flows!

This can make sense because debt and equity markets are not entirely segmented, as suggested previously

Result #3: Valuing PE Fund Portfolios

| | All Deals | Realized Deals |
|------------------|-------------------------|-------------------------|
| CME | 0.335 (0.357) | -0.013 (0.007) |
| $H_0 : CME = 0$ | $t = 0.830$ (0.400) | $t = -1.405$ (0.160) |
| GPME | -0.105 (0.029) | -0.001 (0.002) |
| $H_0 : GPME = 0$ | $t = -3.610$ (0.001) | $t = -0.460$ (0.650) |
| PME | -0.115 (0.025) | -0.017 (0.002) |
| $H_0 : PME = 0$ | $t = -4.590$ (0.000) | $t = -8.690$ (0.000) |

- relative to PME and GPME PE deals underperform on average, and mostly significantly so
- performance looks brighter in light of credit market performance through the lens of CME
 - underperformance may reflect credit market exposure
 - account for such exposure in PE performance evaluation

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A: Debt price data help predict PE performance

Q: Can we use 'credit market equivalent' (CME) valuation to value private firms?

A: Accounting for credit market exposure helps understanding PE valuations