

REDUCED FORM PE-FUND ASSET MODELING

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Agenda

1. Introduction
2. Probability Space
3. Structural vs. Reduced Form Approach (in the PE Context)
4. Exit Dynamics Regression
5. Data & Model Estimation
6. Monte Carlo Simulation
7. Conclusion



1 INTRODUCTION



Title Reduced form private equity fund asset modeling

Status Working paper

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Idea Use reduced form credit risk approach to model the exit behavior of private equity fund investments

Aim Quantify the relationship between exit timing and exit performance on asset level

Application Monte Carlo simulation tool to estimate "Cash-Flow-at-Risks" of private equity portfolios



1.2 PRACTICAL EXAMPLE

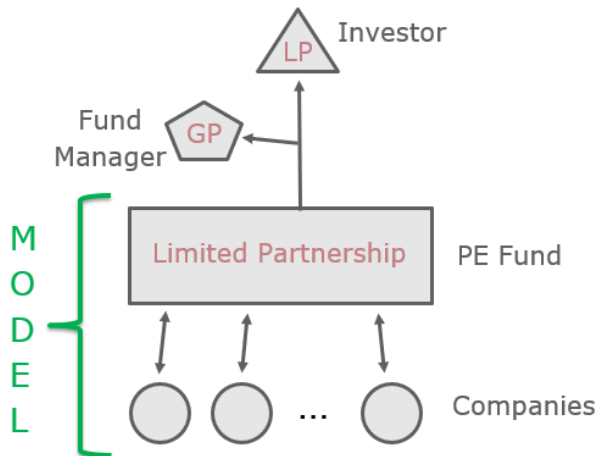
Assume:

- You hold a PE fund consisting of 3 (or 15) underlying companies.
- The fund manager values your share to 1 million €.
- No secondary market for private equity exists.
- You have to wait until the fund manager sells the companies.
- Fund managers charge no fees.

How much money will you receive? When?



1.3 PRIVATE EQUITY FUND STRUCTURE



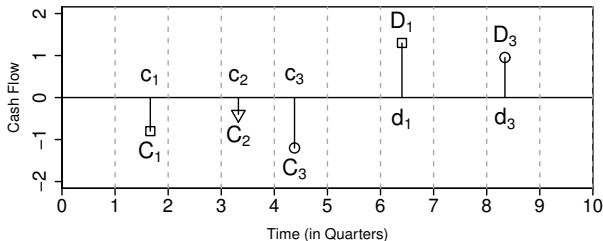
2 PROBABILITY SPACE



2.1 MARKED POINT PROCESS

All fund cash flows are modeled by a marked point process (MPP)

$$(\mathcal{C}_i, \mathcal{D}_i) = (c_i, C_i, d_i, D_i)$$



Event Type	Timing	Amount.
Investment	c_i	$C_i.$
Divestment	d_i	$D_i.$



2.2 AUXILIARY PROCESSES

Let $N^{(d)}(t)$ count all exit events happening before time t .

$$N^{(d)}(t) = \sum_{i=1}^n \mathbf{1}_{\{d_i \leq t\}}$$

In addition there exist 4 auxiliary processes:

Value process $V_i(t) = \hat{V}_i(t) + v_{t,i}$ (proxy for true asset value \hat{V})

Covariate process X_i (public market factors, company details)

Censoring process $Z_i^{(\text{censoring})}$ (indicates active investment)

Filtering process $Z^{(\text{filtering})}$ (due to quarterly reporting in PE)

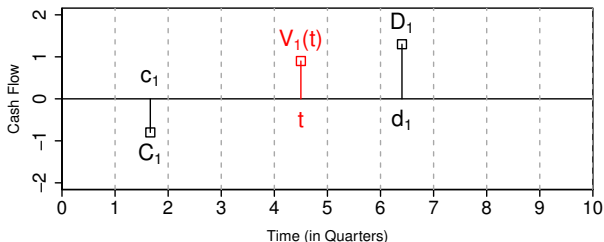


3 STRUCTURAL VS REDUCED FORM



3.1 PRIVATE EQUITY FUND MODEL ON ASSET LEVEL

Definition 1: A private equity fund (PEF) model on asset level connects the intermediate proxy valuation process $V_i(t)$ with the resulting final divestment cash flow \mathcal{D}_i for $t \in (c_i, d_i)$.



Remark The investment part C_i is ignored in our exit model.



3.2 STRUCTURAL PRIVATE EQUITY FUND MODEL

Definition 2: A structural PEF model on asset level is defined in terms of the observable and unobservable firm value processes V and \dot{V}

$$D = C + \int_c^d dV(s) = C + \int_c^d d\dot{V}(s)$$

with

$$d = \inf \{t > c : V(t) \in E(t)\} \wedge \inf \{t > c : \dot{V}(t) \leq 0\}$$

where $E(t)$ is the dynamic exit acceptance set.

Most authors use structural approaches to model PE assets, but ...

- ⚠ Structural models require detailed information.
- ⚠ Incomplete information settings are inherent to private equity.
- ⚠ Scarce data: quarterly reporting, GP appraisals, stale pricing ...



3.3 REDUCED FORM PRIVATE EQUITY FUND MODEL

Definition 3: A reduced form PEF model on asset level is defined in terms of the observable proxy value process V and a stochastic multiplier m_D that is modeled conditional on exit timing, i.e.

$$D = m_D(t|d) \cdot V(t)$$

with

$$d = \inf \{t > c : N^{(d)}(t) > 0\}$$



Reduced form = model based on observable quantities.



The exit timing d is determined by the exit counting process $N^{(d)}$.



$V(t)$ is not perceived as stochastic process, but realized data.



4 EXIT DYNAMICS REGRESSION



Marginal parametric models for the bivariate dependent variable:

$$\text{Timing } y = d - t = m_y(\mathbf{X}|\xi_y) + e_y$$

$$\text{Multiple } Y = \frac{D}{V(t)} = m_Y(\mathbf{X}|\xi_Y, y) + e_Y$$

where ξ_Y, ξ_y are the model parameters
and e_Y, e_y are (possibly mutually dependent) error terms.



Multiple is modeled conditional on **Timing**.



Parametric models are more easy to estimate and simulate.



4.2 TIMING REGRESSION

Intensity based approach


$$\mathbb{E} \left[\int_0^\infty dN_i^{(d)}(u) \right] = \mathbb{E} \left[\int_0^\infty h_i(u | \mathbf{X}(u)) du \right]$$

Multiplicative hazard formulation (a.k.a. Cox Regression)

$$h_i(t | \mathbf{X}(t)) = h_0(t) \exp(\beta \mathbf{X}(t)) Z_i^{(\text{censoring})}(t)$$

Parametric Weibull base hazard

$$H_0^{(\text{wb})}(t | \xi_y) = \int_0^t h_0^{(\text{wb})}(u | \xi_y) du = \left(\frac{t}{\text{scale}_{\text{wb}}} \right)^{\text{shape}_{\text{wb}}}$$

 Stepwise integration of $H^{(\text{wb})}$ due to time-varying covariates.



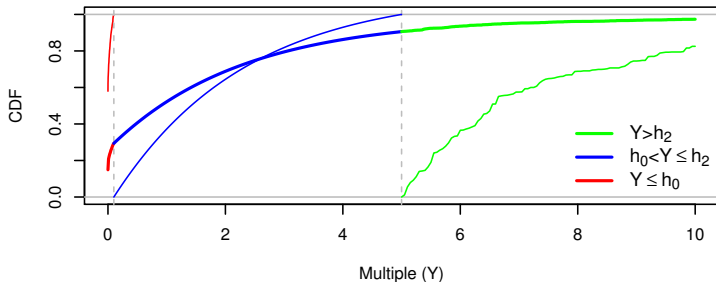
4.3 MULTIPLE REGRESSION

⚠ Private equity multiples are option-like, i.e. many defaults and some very high returns (25% defaults and 10% with return > 400%)

⚠ Normal and log-normal assumption not valid for $Y = \frac{D}{V(t)}$

💡 Truncation: exclude negative and positive outliers

💡 Use three part hurdle model (use empirical CDF in outlier region, Gamma distribution in-between)



5 DATA & MODEL ESTIMATION



5.1 DATA & MODEL ESTIMATION

- Asset level data set of 144 Buy Out and 98 Venture Capital funds
- Both models (Timing & Multiple) are fitted by maximizing the parametric likelihood functions
- Exit Multiple model considerations:
 - Data subset:
 - Exit event before 2017-01-01 (realized investments)
 - Entry event before 2010-01-01 ("quick-flip bias")
 - RVPI-ratio $> 10\%$ (economic relevance)
 - Resampling based estimation approach to account for auto-correlation of quarterly reported net asset values (i.e. sample one observation per company ID in each iteration)



5.2 EXPLANATORY VARIABLES

	Variable name	Timing	Multiple
x_1	Holding period: $t - c$	no	yes
x_2	Time to exit: $d - t$	no	yes (as $x_2 + x_3$)
x_3	Zombie stage: $\max(0, d - c - 10)$	no	yes (as $x_2 + x_3$)
x_4	RVPI: $\frac{V}{C}$	no (internal)	yes
x_5	High yield spread	yes	yes
x_6	Public equity performance	yes	yes
x_7	Fund age at entry (date)	yes	no



5.3 REGRESSION RESULTS (BUY OUT AND VENTURE CAPITAL)

Regression results seem plausible:

- PE-funds exit their companies the faster:
 - the higher the public equity return,
 - the lower the high yield spreads,
 - the older the fund age.
- High Multiples can be expected for:
 - high public equity returns,
 - low high yield spreads,
 - short (future) time-to-exit's.
- Long (past) holding periods decrease the probability of default.
- The lower the RVPI ratio ($\frac{V}{C}$), the higher the Multiple variance.



6 MONTE CARLO SIMULATION



6.1 MONTE CARLO SIMULATION EXAMPLE

Compare two Venture Capital funds with 3 and 15 company holdings:

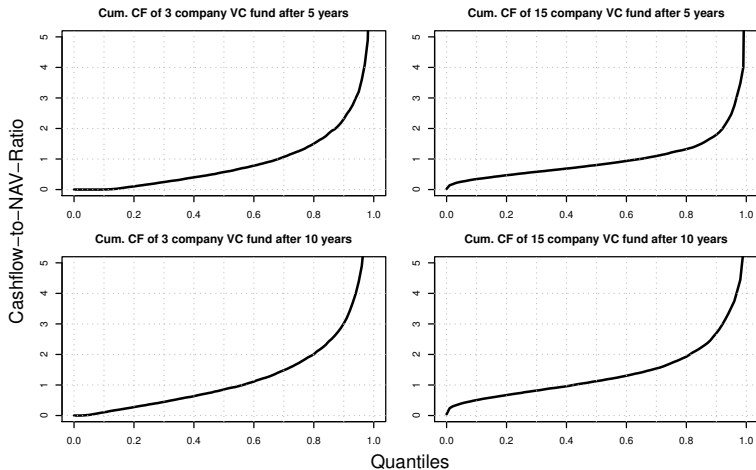
- All companies are already held 5 years and have an RVPI (current value to cost ratio) of 1
- 5,000 simulation iterations
- Fees are not considered
- Public market scenarios are generated by historical simulation

Which fund is riskier?

How much money can you expect after 5 or 10 years (from now)?



6.2 SIMULATION OUTCOME: CASH-FLOW-AT-RISK CHART



6.3 SIMULATION OUTCOME: EXAMPLE QUANTILES

Cash-Flow-at-Risk	3-company	15-company
"10% worst case after 5 years"		
5-year horizon, 10% quantile	0.00	0.34
"10% best case after 10 years"		
10-year horizon, 90% quantile	3.01	2.69

Cash-Flow-at-Risk: Maximum amount of cumulative divestment cash flow you can expect after a given horizon with a given probability (in relation to the current valuation).



7 CONCLUSION



Our paper:

1. contributes a **marked point process** framework to describe the cash flow dynamics of private equity funds on single asset level.
2. studies the **divestment behavior** of private equity funds in a reduced form approach, where the exit multiple is modeled conditional on exit timing.
3. proposes tailored **parametric regression** models, which can include covariates that are infeasible in fund level approaches.
4. highlights the superiority of **reduced form** formulations in incomplete information settings like private equity.



7.2 STRUCTURAL VS REDUCED FORM

General Aim: Model for (d, D) given $V(t)$ and (c, C) .

	Structural	Reduced-form
Stochastic Processes:		
- observable	$V(t)$	$N^{(d)}(t)$
- unobservable	$\dot{V}(t), E(t)$	-
Simulation Procedure:		
1. Step	Multiple paths for $V(t)$ and $\dot{V}(t)$	Timing d directly from survival function
2. Step	Timing check when $E(t)$ is met	Multiple $m_D(t d) \cdot V(t)$



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DO YOU HAVE COMMENTS?

