



# Intangible Assets

Winter 2023

Quantitative Portfolio Management

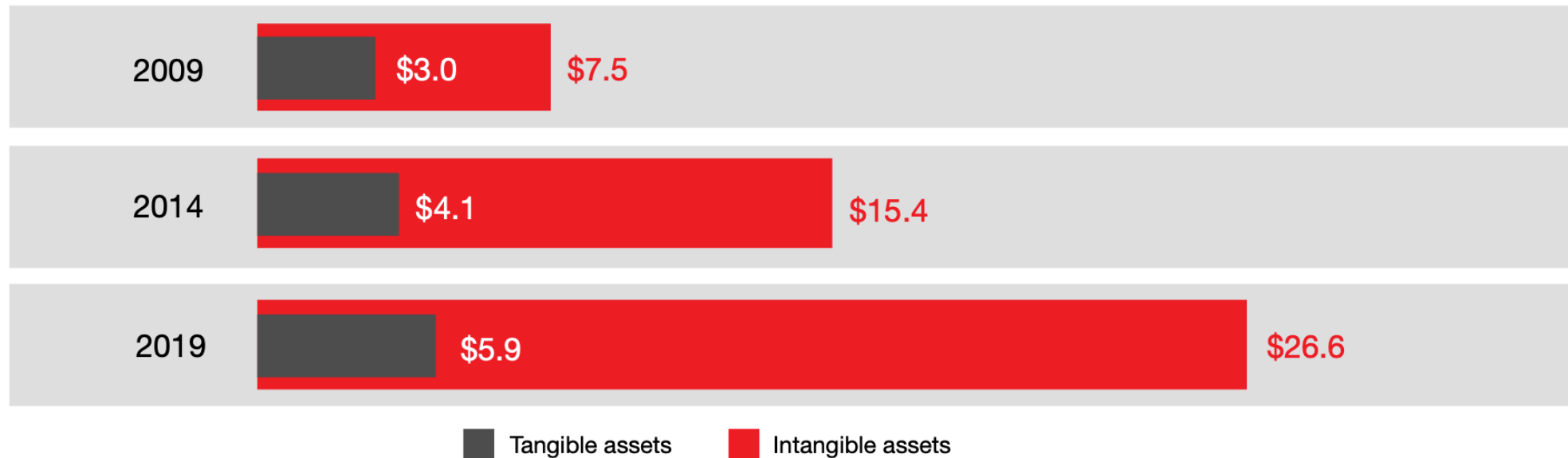
Amy Cai, Arthur Ji, Charene Shen, Anthony Song



# Why Intangibles

- Value accrues to intangible assets has increased significantly in recent years as companies adapt to “information revolution” with virtual business models<sup>(1)</sup>

Values of tangible and intangible assets\* at S&P 500 companies (in trillions)



Source: Company reports, PwC analysis

\*Implied market value of intangibles = Market cap + Book value of total liabilities - Book value of tangibles

# Accounting Rules

- However, accounting rules treat internally developed vs. acquired intangible assets asymmetrically
- Internally generated R&D does not always appear on Balance Sheet and book value of intangible assets (including goodwill) are mostly acquired as a result<sup>(1)</sup>

	IFRS	GAAP
Internally Generated Intangibles	Research costs are immediately expensed / development costs may be capitalized	Not capitalized except for internally developed software
Acquired Intangibles	Capitalized	Capitalized

1. Per PwC Analysis, KPMG Insights

# Main Economic Ideas

- Therefore, we want to test the idea of whether companies / investors tend to overprice these intangible assets during acquisitions, given the risks of change in consumer sentiment, regulation, and technology, which would lower future expected returns and cost of capital
- Examples:
  - 2001 AOL Time Warner merger led to \$99B goodwill impairment in 2002 after the dot-com bubble burst
  - 2016 Valeant Pharmaceuticals under SEC investigation following its aggressive R&D acquisitions
  - 2016 Yahoo! wrote down \$1.1B Tumblr acquisition by \$482 million after it failed to monetize Tumblr effectively through its user base
  - 2019 Kraft Heinz wrote down \$15.4B on Kraft and Oscar Mayer brands due to changing consumer taste

# Economic Interpretations

- **Risk-Based**

- i. Valuation risks of internally developed intangibles as their fair values have not been independently assessed in the active market
- ii. Risks associated with the outcome of the investment (change in market conditions, consumer sentiment, and technological feasibility)
- iii. Risks of infringement by competitors

- **Behavioral**

- i. Acquisition bias as investors might be overly optimistic about the synergies, growth prospects, and intrinsic value associated with acquired intangibles
- ii. Investors may overly focus on accounting metrics and undervalue internally-developed intangibles
- iii. Short-termism

# Connection to Existing Research

## Intangible intensity and stock price crash risk

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School of Finance, Central University of Finance and Economics, China

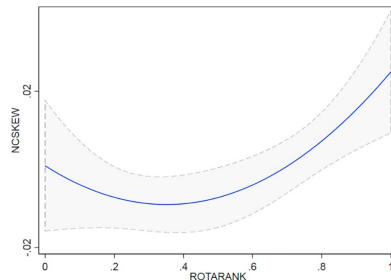


Fig. 2. Nonlinear relationship between intangible intensity and crash risk.

### Variable Definition:

The intangible intensity is decomposed into definite-life intangible intensity FLRANK (a), indefinite-life intangible intensity INFLRANK (b), and goodwill intensity GWRANK (c).

$$\text{ROTA} = \text{EBITDA} / \text{NETPPE} = (a) + (b) + (c)$$

How does intangible intensity affect crash risk?

- Goodwill impairments
- High information asymmetry

## Intangible Value\*

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May 4, 2022

### Signal Construction:

$$B_{it}^{INT} = B_{it} - GDWL_{it} + INT_{it} \rightarrow B_{it}^{INT} / M$$

### Key Findings:

- A value portfolio that **adds intangible capital** to book assets prior to sorting provides **much stronger performance** in all periods.
- Each leg of intangible value outperforms traditional value that is **industry-sorted**.
- Outperformance is more pronounced when **financials, utilities, and public service firms are dropped** during portfolio formation.

The Stock Market Valuation of Research and Development Expenditures

Author(s): Louis K. C. Chan, Josef Lakonishok and Theodore Sougiannis

Source: *The Journal of Finance*, Dec., 2001, Vol. 56, No. 6 (Dec., 2001), pp.

Published by: Wiley for the American Finance Association

### Key Findings:

- Using R&D/sales as signal **does not generate a spread in average returns** or alphas.
- Market is skeptical about firms with high R&D/MarketCap, giving them low valuations.
- The evidence on an association between **R&D intensity measured relative to sales and future returns is not strong**.
- The clearest evidence that high R&D plays a distinctive role arises from stocks with **high R&D relative to the market value of equity**.

# Strategy Construction



Baseline signal:

$$\text{signal} = - \text{intan}$$

- “Intan” represents the value of total intangible asset on a company’s balance sheet
- Negative sign captures our main economic idea: penalize companies with high intangible assets

Control industry:

**Technology sector**

- Different industries may have different number of patents, copyright, etc. on average
- Make apple-to-apple comparison within only one industry for intangible assets

Control size:

$$\text{Signal} = \frac{- \text{Intan}}{\text{at}}$$

- Larger firms may have greater number of intangible assets
- Add denominator “at”, representing total assets, to our signal to control for different firm size

Adjust current asset:

$$\text{Signal} = \frac{- \text{Intan}}{\text{at} - \text{act}}$$

- Current assets such as account receivable, etc. is frequently changed by daily operations
- Exclude “act”, representing total current asset, from our signal denominator

# Strategy Construction (cont'd)

However, it could also be possible that the market has priced in our economic idea: those tech firms with proportionally high intangible assets has already been penalized by a discounted valuation

To compensate for potentially repeated penalization, we add a **value signal** =  $\frac{at - lt}{mkvalt} \left( \frac{\text{book equity value}}{\text{market capitalization}} \right)$

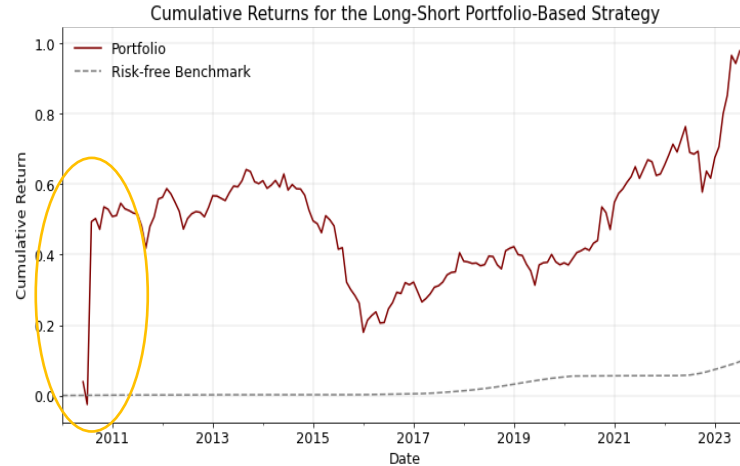
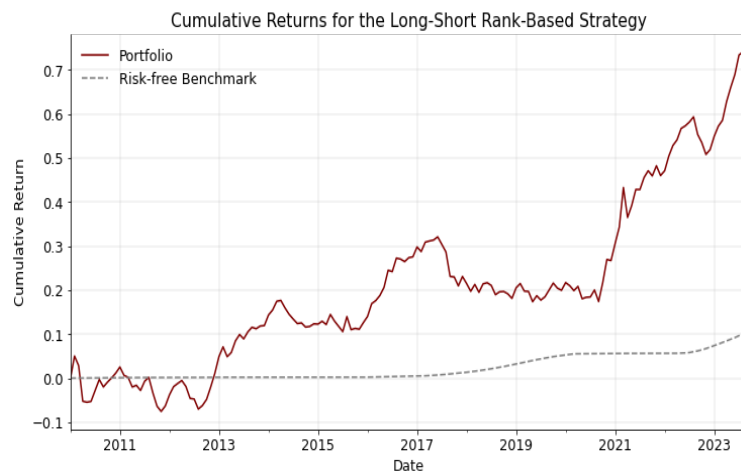
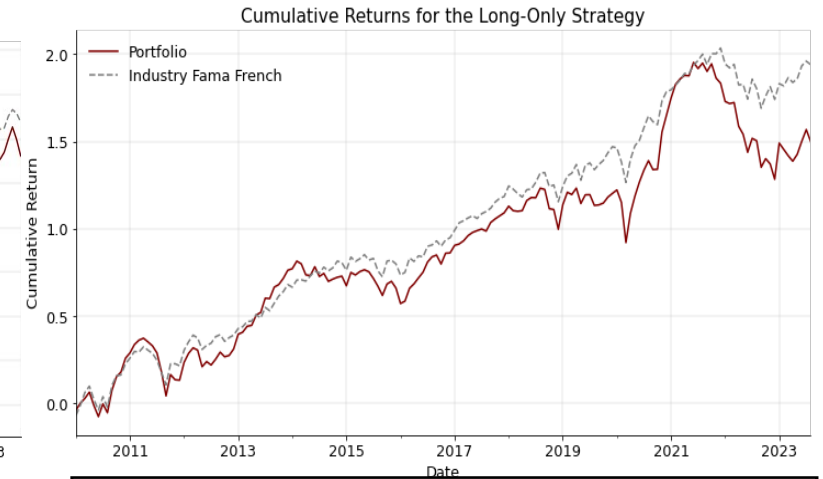
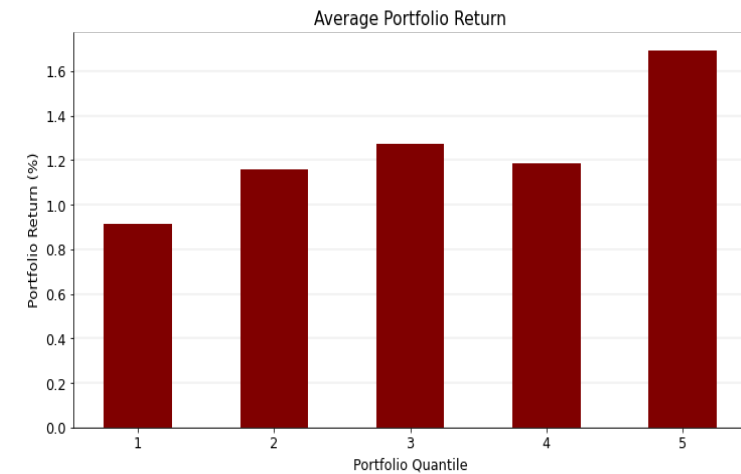
**Combined signal** =  $50\% \frac{-Intan}{at-act} + 50\% \frac{at - lt}{mkvalt}$  (after normalizing signal value to z-score)

- ➔ High intangible asset, high current valuation: penalize the most
- ➔ High intangible asset, low current valuation: fine, has been penalized
- ➔ Low intangible asset, high current valuation: fine, fairly priced
- ➔ Low intangible asset, low current valuation: reward the most





# Result: Combined Signal



## Observations on Performances:

- Positive relationship between the quintile rank and portfolio performance (Fig 1)
- Outperform risk-free benchmark
- Periods of outperformance and underperformance compared to FF industry portfolios

# Result: Combined Signal (cont'd)

> Running Factor Regressions: Table 1 - 3 Fama-French Factors

	(1)	(2)	(3)	(4)	(5)	(6)
const	0.0084 (0.0051)	-0.0030 (0.0025)	0.0042** (0.0017)	0.0086 (0.0052)	-0.0007 (0.0018)	0.0049*** (0.0017)
mktrf	-0.1208 (0.1137)	1.2845*** (0.0550)	0.0257 (0.0382)	-0.1435 (0.1229)	1.0980*** (0.0429)	-0.0054 (0.0401)
hml				-0.0900 (0.1568)	-0.2133*** (0.0547)	0.1016** (0.0512)
smb				0.1227 (0.2151)	0.9031*** (0.0751)	0.1233* (0.0702)
R-squared	0.0071	0.7752	0.0029	0.0104	0.8836	0.0602
R-squared Adj.	0.0008	0.7737	-0.0035	-0.0087	0.8814	0.0421
N	160	160	160	160	160	160
R2	0.01	0.78	0.00	0.01	0.88	0.06

Standard errors in parentheses.

\* p<.1, \*\* p<.05, \*\*\*p<.01

(1): Long-Short Value Weights ~ CAPM Model

(2): Long-Only Rank Weights ~ CAPM Model

(3): Long-Short Rank Weights ~ CAPM Model

(4): Long-Short Value Weights ~ 3-Factor Fama French Model

(5): Long-Only Rank Weights ~ 3-Factor Fama French Model

(6): Long-Short Rank Weights ~ 3-Factor Fama French Model

Annualized Information Ratios:

	1	2	3	4	5	6
Alpha	0.101	-0.036	0.051	0.103	-0.009	0.058
Std(resid)	0.217	0.105	0.073	0.217	0.076	0.071
Information Ratio	0.464	-0.345	0.700	0.474	-0.114	0.826

> Running Factor Regressions: Table 2 - 5 Fama-French Factors + Momentum

	(1)	(2)	(3)	(4)	(5)	(6)
const	0.0099* (0.0053)	0.0008 (0.0017)	0.0047*** (0.0017)	0.0108** (0.0052)	0.0015 (0.0016)	0.0053*** (0.0017)
mktrf	-0.1529 (0.1257)	1.1153*** (0.0403)	-0.0020 (0.0412)	-0.2217* (0.1294)	1.0658*** (0.0393)	-0.0461 (0.0409)
hml	0.0965 (0.2133)	-0.0881 (0.0684)	0.0725 (0.0700)	-0.0449 (0.2232)	-0.1898*** (0.0678)	-0.0180 (0.0706)
smb	-0.0090 (0.2437)	0.6994*** (0.0781)	0.1349* (0.0800)	-0.0664 (0.2432)	0.6582*** (0.0739)	0.0982 (0.0769)
rmw	-0.2542 (0.2886)	-0.4934*** (0.0925)	0.0145 (0.0947)	-0.2911 (0.2865)	-0.5199*** (0.0870)	-0.0091 (0.0906)
cma	-0.3468 (0.3178)	-0.1391 (0.1019)	0.0615 (0.1043)	-0.1937 (0.3244)	-0.0290 (0.0985)	0.1595 (0.1026)
umd				-0.3180* (0.1615)	-0.2287*** (0.0491)	-0.2035*** (0.0511)
R-squared	0.0232	0.9030	0.0625	0.0474	0.9151	0.1506
R-squared Adj.	-0.0085	0.8998	0.0320	0.0100	0.9117	0.1173
N	160	160	160	160	160	160
R2	0.02	0.90	0.06	0.05	0.92	0.15

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Annualized Information Ratios:

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Alpha	0.119	0.010	0.057	0.130	0.018	0.064
Std(resid)	0.215	0.069	0.071	0.213	0.065	0.067
Information Ratio	0.551	0.143	0.801	0.611	0.279	0.949

## Observations on Alphas:

- increases and becomes more significant as more factors are included: performance is unexplained when adding more factors
- most significant for rank-based portfolios and long-short portfolios
  - These portfolios use signal more effectively
  - long-short strategies fit our economic idea more, because a high ratio is viewed as a notably adverse signal, outweighing the positive implications of a low ratio.

# Result: Combined Signal (cont'd)

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R-squared Adj.	-0.0085	0.8998	0.0320	0.0100	0.9117	0.1173
N	160	160	160	160	160	160
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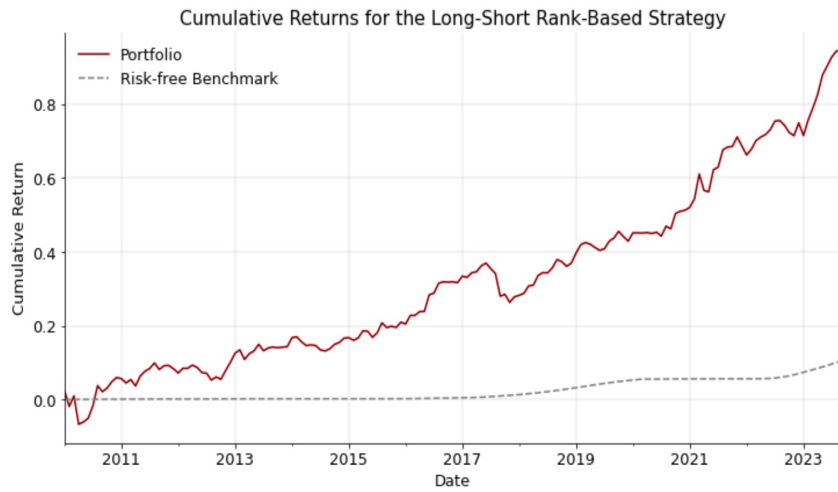
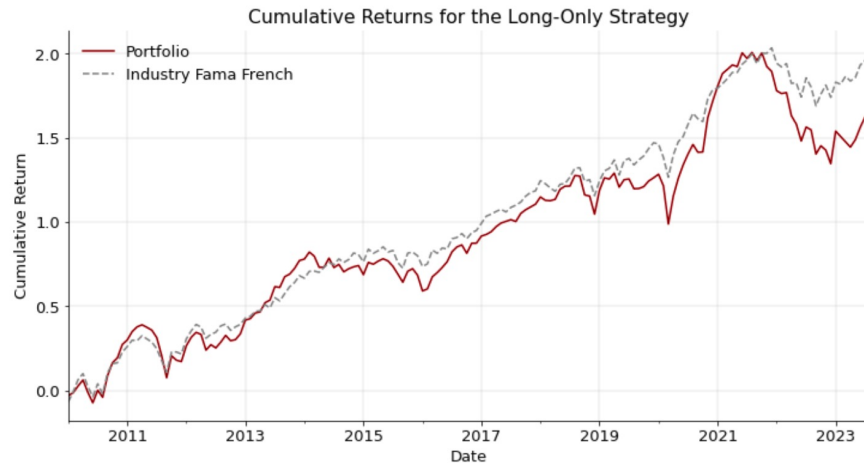
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Std(resid)	0.215	0.069	0.071	0.213	0.065	0.067
Information Ratio	0.551	0.143	0.801	0.611	0.279	0.949

## Observations on Betas:

- By the left figure, the betas of CAPM and 3-Factor FF model are more significant for rank-based strategy (2), (3), (5), (6).
- Positive and significant SMB for rank-based strategy (2), (3), (5) shows it prefers smaller companies.
- Consistently negative and significant UMD shows it potentially performs better during market downturns.
- There is not much consistent patterns in the Betas, which suggests factors have limited explanatory power regarding the signal performance.

# Comparison: Without Value Signal



> Running Factor Regressions: Table 1 - 3 Fama-French Factors

	(1)	(2)	(3)	(4)	(5)	(6)
const	0.0057* (0.0031)	-0.0026 (0.0024)	0.0061*** (0.0015)	0.0055* (0.0031)	-0.0004 (0.0017)	0.0060*** (0.0015)
mktrf	0.0435 (0.0684)	1.2713*** (0.0531)	-0.0271 (0.0334)	0.0368 (0.0727)	1.0904*** (0.0403)	-0.0359 (0.0356)
hml				-0.2217** (0.0928)	-0.2623*** (0.0514)	-0.0942** (0.0454)
smb				0.0745 (0.1273)	0.8870*** (0.0706)	0.0589 (0.0623)
R-squared	0.0025	0.7839	0.0042	0.0378	0.8938	0.0319
R-squared Adj.	-0.0038	0.7825	-0.0021	0.0193	0.8917	0.0133
N	160	160	160	160	160	160
R2	0.00	0.78	0.00	0.04	0.89	0.03

Standard errors in parentheses.

\* p<.1, \*\* p<.05, \*\*\*p<.01

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(5): Long-Only Rank Weights ~ 3-Factor Fama French Model

(6): Long-Short Rank Weights ~ 3-Factor Fama French Model

Annualized Information Ratios:

	1	2	3	4	5	6
Alpha	0.069	-0.031	0.073	0.065	-0.005	0.073
Std(resid)	0.131	0.101	0.064	0.128	0.071	0.063
Information Ratio	0.528	-0.303	1.147	0.511	-0.072	1.156

## Observations on Performances:

- Higher alpha for single-signal strategy
- Beta still inconsistent

## Potential explanations:

- Traditional value metrics (e.g., B/M ratio) cancel out certain intangible signal
- B/M is not an ideal valuation factor due to the high volatility of the technology sector
- Timing of the data

# Implications for Further Research

1. Use the strategy in other sectors
2. more focusd on accounting quality (z score, f score, accruals – CFO) and valuation risks