

# Asset Similarity, CO2 Emissions, and Mutual Fund Performance

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## Era il 2012

## COMPARATIVE RETURNS

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Range 4/30/90 - 12/30/11 Period Monthly 260 Mo. Period

Securities	Crcncy	Prc Appr	Total Ret	Difference	Annual Eq
1 TKLD400U Index	USD	578.08 %	578.08 %*	77.47 %	9.23 %
2 SPX Index	USD	280.17 %	500.61 %		8.62 %
3					

(\* = No dividends or coupons)



## Mutual funds: Stylized facts

Delegated portfolio management industry plays a crucial role in financial markets. According to the Investment Company Institute, in fact, at year-end 2021:

- ▶ Total worldwide assets invested in regulated open-end funds reached 71.1 trillion of US dollars. US open-end funds (8,887 mutual funds and 2,690 ETFs) account for 48.1% of such wealth;
- ▶ In particular, the US mutual fund industry remained the largest in the world with 27 trillion in total net assets (vs. 150 billion at the end of '80s);
- ▶ Almost 45% of US households invested in mutual funds (only 5% in 1980), with an aggregate investment of over 23 trillion dollars;
- ▶ Equity mutual funds represented 55% of the overall US mutual funds industry, and about 90% of them were actively managed.

## Motivation: Average performance and Tails

Despite its importance, the added value of active investment management remains a long-standing controversy:

- ▶ There is a broad consensus that, on average, open end equity mutual funds do not outperform the stock market (Fama and French, 2010, Carhart, 1997, Ippolito, 1989, and Sharpe, 1991);
- ▶ However, actively managed open end equity mutual funds exhibit a considerable cross-sectional variation, with only a few able to generate positive risk-adjusted returns. In particular, successful funds exhibit different investment characteristics/behaviors:
  - concentrate their portfolios in industries** where they have informational advantages (Kacperczyk, Sialm, and Zheng, 2005); **rely less on public information** (Kacperczyk and Seru, 2007); **focus on stock picking and market timing strategies** according to the state of the economy (Kacperczyk, Nieuwerburgh, and Veldkamp, 2014); **trade more** (Daniel, Grinblatt, Titman, and Wermers, 1997; Chen, Jegadeesh, and Wermers, 2000; Pástor, Stambaugh, and Taylor, 2017); **have less wealth to manage** (Chen, Hong, Huang, and Kubik, 2004); **deviate more from benchmark portfolios** (Cremers and Petajisto, 2009) and **from the decisions of their predecessors** (Jiang and Verardo, 2018).

# Our Contribution I

- In this paper we investigate the cross-sectional distribution of skill (risk-adjusted returns) among actively managed open end equity mutual funds.

We uncover a novel investment characteristic of successful funds based on the degree to which fund holdings do not overlap with the holdings of other funds (**asset similarity**).

- Our estimates reveal a large degree of heterogeneity in asset similarity across funds, with some funds exhibiting a tendency to follow the crowd while others show a propensity to hold 'unique' portfolios.

We find that differences in asset similarity across funds *predict* mutual fund extra performance both at portfolio and fund levels:

- ① Funds with low similarity exhibit positive risk-adjusted gross returns of roughly 1.3% per year. By contrast, high-similarity funds do not exhibit any significant extra performance;
- ② In multivariate predictive regressions, asset similarity can predict four-factor alphas after controlling for funds characteristics (such as fund size, age, turnover, expense ratios, flows) and holdings characteristics (such as market cap, growth opportunities, momentum, investment and profitability);
- ③ The negative relation between the fund performance and its degree of similarity is persistent over time, with risk-adjusted returns that are large and significant over horizons of up to one year after the measurement of fund similarity. This result suggests that the link between asset similarity and future performance is not due to chance.

## Our Contribution II

- In addition to predictability of mutual fund extra performance, we test whether low-similarity funds consistently make better investment decisions than high-similarity funds. Our results indicate that the former systematically increase(decrease) their exposure in undervalued(overvalued) stocks, that is they exhibit higher (*chasing alpha*) ability;
- Finally, we also investigate how changes in *fund management structure* affect fund similarity. Specifically, our guess is that fund similarity depends on new management hires and leaves, according to the specific characteristics of the incoming manager's fund.

All results are robust to several asset pricing models, and other investment characteristics (such as herding behavior, and industry concentration).

## Data Sources

- ▶ Time Horizon: monthly observations from 12/2005 to 06/2018 (151 months, 51 quarters) [We already have data up to July 2023];
- ▶ Unique Open-End Equity Mutual Funds: 1,678 (Morningstar - DIRECT);
- ▶ Unique Holdings: 7,037 (Morningstar - EDW);
- ▶ CRSP/COMPUSTAT:
 

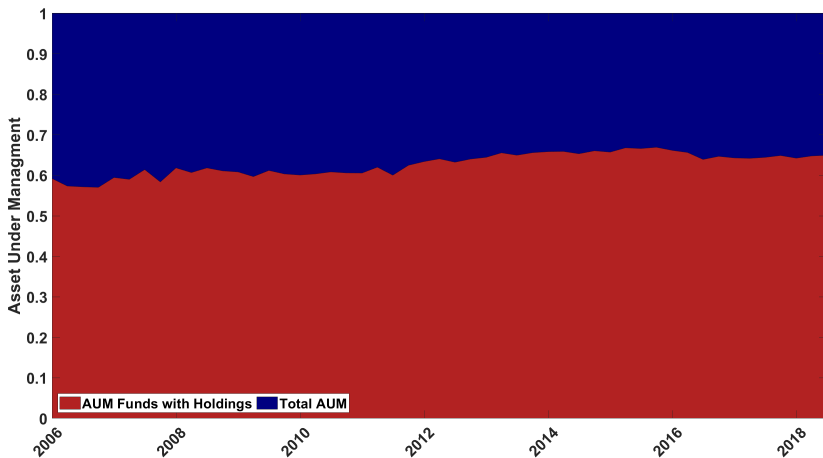
Price and Dividend; Market Value of Equity; Book Value of Equity; Total Assets; Net Sales or Revenues; Selling General and Administrative Expenses; Interest Expense on Debt; and Cost of Goods Sold. These variables are used to create size ( $ME$ ), book-to-market ( $BtM$ ), Investment ( $Inv$ ), Operating Profitability ( $OP$ ), and momentum characteristics ( $Ret_{-11}$ ) following Fama and French (2012, 2017) procedures.
- ▶ Refinitiv:
 

Scope 1 and Scope 2 emissions used to compute the total Green House Gasses ( $GHG$ ) emission (with 669 unique holdings of 7,037). The environmental score ( $E$ , with 2,084 unique holdings of 7,037).
- ▶ North American Risk-Factors: [Fama-French Website](#) (Fama and French, 2012, 2017)
 

Market Excess Return ( $R_m^e$ ); Small minus Big ( $SMB$ ); High minus Low ( $HML$ ); Momentum ( $MoM$ ); Robust minus Weak ( $RML$ ); and Conservative Minus Aggressive ( $CMA$ ).



# Market Coverage



## Asset Similarity: Cosine Similarity Measure

We measure the portfolio overlap between two funds using cosine similarity at the issuer level. Specifically, the cosine similarity between the portfolios of fund  $i$  and  $j$  at quarter-end  $t$  is the dot product of the pair's portfolio weight vectors normalized by the vectors' lengths, that is:

$$\text{Similarity}_{i,j,t} = \frac{W_{i,t} \cdot W_{j,t}}{\|W_{i,t}\| \cdot \|W_{j,t}\|} \quad (1)$$

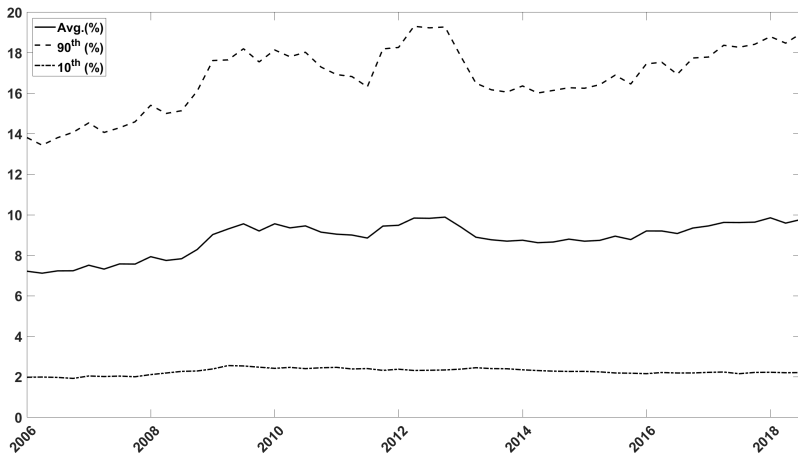
where  $W_{i,t}$  and  $W_{j,t}$  are funds  $i$  and  $j$  vector of weights at quarter-end  $t$ , respectively.

Because all portfolio weight vectors have non-negative elements, cosine similarity is bounded in the interval  $[0,1]$ . Precisely:

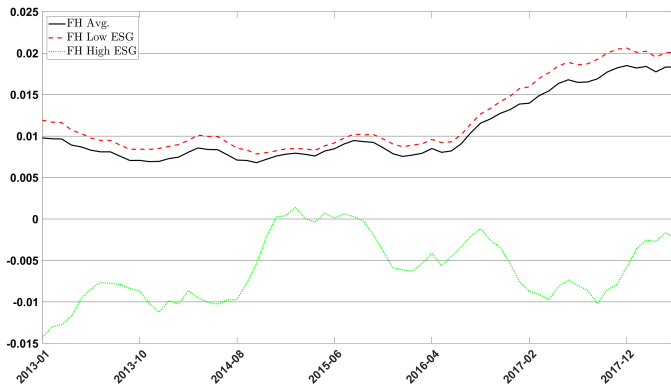
- ▶  $\text{Similarity}_{i,j,t} = 0$  if the portfolios of funds  $i$  and  $j$  are completely different;
- ▶  $\text{Similarity}_{i,j,t} = 1$  if the portfolios of funds  $i$  and  $j$  are exactly the same,

as Girardi et al. (2021) in the case of the insurance market, and Sias et al. (2016) in the case of hedge funds.

# Asset Similarity



# Fund Hearing...from SRI 2020



# Summary Statistics

Our Summary statistics confirm those of Jiang and Verardo (2018)

	Mean	Std.Dev.	10 <sup>th</sup> Pctl	50 <sup>th</sup> Pctl	90 <sup>th</sup> Pctl	N.of Obs
<i>Gross Returns (%)</i>	2.58	8.66	-9.01	3.64	12.11	69128
<i>Fund Size (mln)</i>	1645.17	6327.99	29.68	324.40	3327.33	76489
<i>Flow (%)</i>	2.64	18.89	-13.06	0.99	17.10	77194
<i>Expense (%)</i>	1.34	0.50	0.82	1.30	1.85	18836
<i>Turnover (%)</i>	79.29	261.35	14.00	53.00	150.00	17682
<i>Fund Age (years)</i>	9.76	9.37	1.25	7.75	19.25	158331

▶ Holding-Level Summary Statistics

▶ Variables Definition

# Determinants of Fund Similarity

	(1)	(2)	(3)	(4)
<i>Log(Fund Size)</i>	0.000 (0.002)	0.000 (0.002)	-0.001 (0.001)	-0.001 (0.001)
<i>Log(Fund Age)</i>	0.001 (0.002)	0.001 (0.002)	-0.001 (0.001)	-0.001 (0.001)
<i>Expense</i>	-0.001 (0.001)	-0.001 (0.001)	0.000 (0.001)	0.000 (0.001)
<i>Flow</i>	0.001*** (0.000)	0.001*** (0.000)	0.000*** (0.000)	0.000*** (0.000)
<i>Turnover</i>	-0.002*** (0.000)	-0.002*** (0.001)	-0.000 (0.000)	-0.001 (0.000)
<i>ICI</i>		-0.012*** (0.002)	-0.007*** (0.002)	-0.007*** (0.002)
<i>FH</i>		0.001*** (0.000)	0.000 (0.000)	0.000 (0.000)
<i>ME</i>			0.063*** (0.005)	0.060*** (0.005)
<i>BtM</i>			0.001* (0.001)	0.001 (0.001)
<i>INV</i>			0.000* (0.000)	0.000* (0.000)
<i>OP</i>			-0.000 (0.000)	-0.000 (0.000)
<i>Ret<sub>-11</sub></i>			0.002*** (0.000)	0.002*** (0.000)
<i>Log(GHG)</i>				0.003*** (0.001)
<i>Constant</i>	0.085*** (0.003)	0.088*** (0.003)	0.080*** (0.002)	0.081*** (0.002)
<i>N.ofObs</i>	110850.000	105510.000	92916.000	92916.000
<i>R<sup>2</sup><sub>Adj</sub></i>	0.030	0.062	0.726	0.732

## Additional Summary Statistics on Similarity-based portfolios

	<i>Low</i>	2	3	4	5	6	7	8	9	<i>High</i>
<i>Similarity</i>	1.30	2.35	3.22	4.53	6.76	9.70	12.49	15.03	17.57	21.79
<i>Fund Size</i>	484.30	886.25	1265.48	1569.49	1473.38	1485.61	1680.37	2764.94	4420.11	2603.78
<i>Flow</i>	1.74	2.35	2.44	2.52	2.73	3.48	3.00	2.26	2.08	2.03
<i>Expense</i>	1.53	1.43	1.35	1.31	1.36	1.36	1.33	1.29	1.23	1.17
<i>Turnover</i>	62.84	73.20	84.15	85.05	95.30	82.88	70.29	69.69	67.65	93.77
<i>Fund Age</i>	10.87	9.03	8.70	9.38	10.73	11.06	10.91	9.46	10.15	9.34
<i>log(GHG)</i>	14.72	15.32	15.39	15.42	15.41	15.44	15.45	15.45	15.44	15.42
<i>Avg. Gross Returns (%)</i>	0.92	0.93	0.91	0.89	0.86	0.81	0.81	0.80	0.81	0.80
<i>Std. Gross Returns (%)</i>	4.98	4.83	4.66	4.52	4.37	4.24	4.16	4.14	4.15	4.10
<i>Sharpe Ratio</i>	0.18	0.19	0.19	0.19	0.19	0.19	0.19	0.20	0.20	0.19
<i>N. of Months</i>	151	151	151	151	151	151	151	151	151	151

# Predictability: Asset Similarity and Performance across portfolios

	<i>Low</i>	2	3	4	5	6	7	8	9	<i>High</i>	<i>L-H</i>
<b>Gross Returns</b>											
FF $\hat{\alpha}$	0.121*** (0.044)	0.111** (0.042)	0.096** (0.042)	0.077* (0.046)	0.039 (0.043)	0.006 (0.036)	0.013 (0.031)	0.004 (0.028)	0.013 (0.037)	0.016 (0.029)	0.105** (0.052)
Carhart $\hat{\alpha}$	0.124*** (0.043)	0.11** (0.041)	0.093** (0.042)	0.076* (0.046)	0.041 (0.043)	0.007 (0.037)	0.014 (0.031)	0.003 (0.028)	0.012 (0.037)	0.014 (0.029)	0.11** (0.05)
FF5 $\hat{\alpha}$	0.122*** (0.043)	0.105** (0.042)	0.095** (0.044)	0.079* (0.048)	0.066 (0.041)	0.026 (0.036)	0.026 (0.032)	0.001 (0.03)	0.015 (0.039)	0.007 (0.03)	0.116** (0.052)
FFL $\hat{\alpha}$	0.124*** (0.043)	0.107** (0.042)	0.094** (0.042)	0.079* (0.046)	0.043 (0.042)	0.011 (0.036)	0.014 (0.031)	0.003 (0.029)	0.008 (0.036)	0.012 (0.028)	0.112** (0.051)
<i>N. of Months</i>	151	151	151	151	151	151	151	151	151	151	151
<b>Net Returns</b>											
FF $\hat{\alpha}$	-0.006 (0.044)	-0.007 (0.042)	-0.016 (0.042)	-0.032 (0.046)	-0.073* (0.043)	-0.107*** (0.036)	-0.097*** (0.031)	-0.103*** (0.028)	-0.089** (0.037)	-0.081** (0.029)	0.075 (0.052)
Carhart $\hat{\alpha}$	-0.003 (0.043)	-0.009 (0.041)	-0.019 (0.042)	-0.033 (0.046)	-0.071* (0.043)	-0.106*** (0.037)	-0.096*** (0.031)	-0.104*** (0.028)	-0.09** (0.037)	-0.083*** (0.029)	0.08* (0.05)
FF5 $\hat{\alpha}$	-0.004 (0.043)	-0.014 (0.042)	-0.016 (0.044)	-0.03 (0.047)	-0.046 (0.041)	-0.087** (0.036)	-0.084** (0.032)	-0.106*** (0.03)	-0.087** (0.039)	-0.09*** (0.03)	0.086* (0.052)
FFL $\hat{\alpha}$	-0.003 (0.043)	-0.011 (0.042)	-0.018 (0.042)	-0.029 (0.046)	-0.069* (0.042)	-0.103*** (0.036)	-0.096*** (0.031)	-0.104*** (0.029)	-0.094** (0.036)	-0.086*** (0.028)	0.082* (0.051)
<i>N. of Months</i>	151	151	151	151	151	151	151	151	151	151	151



# Conclusions

- In this paper we investigate the cross-sectional distribution of risk-adjusted returns among actively managed open end equity mutual funds;
- We uncover a novel investment characteristic of successful managers based on the degree to which the fund holdings do not overlap with the holdings of other funds (**asset similarity**):
  - Specifically, we find that differences in asset similarity across funds *predict* mutual fund extra performance: funds with low similarity exhibit a positive risk-adjusted performance which persists over time;
  - The ability of low similarity funds to beat the market strongly depends on their ability to buy(sell) undervalued(overvalued) stocks consistently (*Chasing Alpha*);
- (Our guess is that...) changes in *fund management structure* affect fund similarity.

# Conclusions

**Thank you for Your Attention!**

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- Mutual funds manager's skills: Cohen, Coval, and Pástor (2005), Kacperczyk, Nieuwerburgh, and Veldkamp (2014), Berk and Van Binsbergen (2015), Barber, Huang, and Odean (2016), Jiang and Verardo (2018)
- Fund performance and the business cycle: Ferson and Schadt (1996), Christopherson, Ferson, and Glassman (1998), and Moskowitz (2000).
- Fund turnover and performance: Wermers (2000), Kacperczyk, Sialm, and Zheng (2005), and Edelen, Evans, and Kadlec (2007) find no significant relation, Carhart (1997) find a negative relation and Dahlquist, Engström, and Söderlind (2000), Chen, Jegadeesh, and Wermers (2000) and Pástor, Stambaugh, and Taylor (2017) find a positive relation.
- Asset-similarity: Girardi, Hanley, Nikolova, Pelizzon, and Sherman (2021), Sias, Turtle, and Zykaj (2016)

# Holding-Level Summary Statistics

	<i>Mean</i>	<i>Std.</i>	<i>10<sup>th</sup></i>	<i>50<sup>th</sup></i>	<i>90<sup>th</sup></i>	<i>N. of Obs.</i>
Panel A: Investment characteristics						
<i>ME</i>	4452.24	18830.41	44.89	545.40	7802.20	594556
<i>BtM</i>	0.74	0.88	0.18	0.56	1.36	530830
<i>Inv</i>	17.21	55.02	-14.10	5.41	48.41	605400
<i>OP</i>	17.24	50.19	-18.03	17.77	50.40	581665
<i>Ret – 11</i>	10.46	48.73	-41.01	10.70	58.10	542457
<i>Log(GHG)</i>	13.28	2.13	10.69	13.16	16.28	41650
Panel B: Industries weights across funds						
<i>Business equip. and serv.</i>	22.95	1.11	21.59	22.82	24.36	1398703
<i>Consumer durables</i>	3.22	0.23	2.94	3.19	3.61	228986
<i>Consumer non-durables</i>	6.81	0.29	6.42	6.85	7.21	463591
<i>Energy</i>	6.86	1.32	5.07	7.10	8.36	359238
<i>Finance</i>	19.04	1.80	17.11	18.98	21.88	1322677
<i>Healthcare</i>	10.98	0.73	10.13	10.92	11.76	675443
<i>Manufacturing</i>	13.71	0.80	12.56	13.70	14.82	1070250
<i>Telecom</i>	3.06	0.23	2.79	3.07	3.31	159659
<i>Utilities</i>	3.33	0.35	2.87	3.31	3.88	231056
<i>Wholesale and retail</i>	10.04	0.63	9.13	10.02	10.78	686665

[▶ Back Funds Summary](#)

[▶ Variables Definition](#)



# Holding Level Summary Statistics for Similarity-based Portfolios

	<i>Low</i>	<i>2</i>	<i>3</i>	<i>4</i>	<i>5</i>	<i>6</i>	<i>7</i>	<i>8</i>	<i>9</i>	<i>High</i>
<b>Panel A: Investment characteristics</b>										
<i>ME</i>	1841.75	3036.31	4349.38	5215.05	6808.64	11470.63	12516.11	12681.43	11289.68	6809.93
<i>BtM</i>	0.71	0.71	0.71	0.70	0.61	0.54	0.55	0.54	0.56	0.63
<i>Inv</i>	16.47	16.69	15.97	15.68	16.67	17.71	15.97	16.36	16.15	14.89
<i>OP</i>	20.89	21.93	22.35	21.61	24.19	28.44	30.62	30.71	29.86	25.34
<i>Ret – 11</i>	12.85	12.68	12.63	12.42	13.06	13.57	13.44	13.87	13.98	12.60
<b>Panel B: Industries weights across funds</b>										
<i>Business equip. and serv.</i>	23.07	21.54	20.74	21.94	21.81	20.20	22.79	22.02	21.54	22.70
<i>Consumer durables</i>	2.99	3.18	3.50	3.10	3.49	3.49	3.35	3.78	3.20	3.14
<i>Consumer non-durables</i>	7.04	6.25	7.10	7.31	6.97	7.57	7.93	7.21	7.25	7.20
<i>Energy</i>	3.77	3.73	4.05	4.43	4.16	4.35	5.13	5.16	4.42	4.05
<i>Finance</i>	25.61	24.66	24.36	23.41	22.31	22.83	20.19	19.13	20.67	21.44
<i>Healthcare</i>	9.28	9.74	9.64	10.63	9.80	10.35	8.27	8.60	9.01	9.35
<i>Manufacturing</i>	15.55	16.97	16.15	15.23	15.96	17.09	16.33	17.48	16.85	16.58
<i>Telecom</i>	1.68	1.88	2.53	2.26	2.80	2.22	2.00	2.20	2.27	2.25
<i>Utilities</i>	2.50	2.88	2.73	2.63	2.96	3.85	4.79	4.37	3.60	3.58
<i>Wholesale and retail</i>	8.52	9.19	9.20	9.06	9.75	8.05	9.20	10.05	11.19	9.71

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## Variables Definition

Returns: Fund returns, in percentage;

Fund Size: Assets Under Management, in millions of dollars;

Flow: Assets Under Management growth rate, in percentage;

Expense: is the yearly Expense Ratio that includes operating expenses and management fees, including 12b-1 fees, administrative fees, and all other asset-based costs, in percentage;

Turnover: is the percentage of a fund's holdings that have changed over the past year;

Fund Age: Time since fund inception, in years;

Similarity: Fund similarity computed as in equation (1), in percentage;

ME: Value-weighted market capitalization of the holdings, in millions of dollars;

BtM: Value-weighted book-to-market ratio of the holdings, in percentage;

Inv: Value-weighted investment growth ratio of holdings, in percentage;

OP: Value-weighted operating profitability ratio of the holdings, in percentage;

Ret-11: Value-weighted momentum of the holdings, in percentage;

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