Seminario sobre Fondos de Inversión

Vniver§itatö́ dValència

SMART MONEY: A further look at investors' abilities



Brief revision to prior literature

Looking for answers to two relevant questions:

- How do investors select among funds?
 - Ippolito (1992) and Sirri and Tufano (1998), among others, highlight the importance of past performance
- Are they able to anticipate superior returns? This question is still unsolved given the different conclusions observed.

Brief revision to prior literature (II): Different conclusions on Smart Money

- Seminal papers find this phenomenon... Gruber (1996) and Zheng (1999) conclude that investor anticipate fund returns
- □ ... but recent papers do not
 - Ke *et al.* (2005) and Braverman *et al.* (2007) say that fund investors are bad performers.
 - Sapp and Tiwari (2004) indicate that seminal papers are biased by momentum

Our study

All Spanish domestic equity funds
Free of survivorship bias. 240 funds.

□ From January 1999 to December 2006

Monthly data of TNA and investors as well as monthly data of money and investor flows

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Monthly data of TNA and investors as well as monthly data of money and investor flows This is the first study that analyses investor abilities

Our study

All Spanish domestic equity funds from January 1999 to December 2006

Monthly data of TNA and investors as well as monthly data of money and investor flows

Separate data of inflows and outflows

Our study

All Spanish domestic equity funds from January 1999 to December 2006

Monthly data of TNA and investors as well as monthly data of money and investor flows

Separate data of inflows and outflows

Only Keswani and Stolin (2008) have collected a similar dataset, providing evidence of smart purchases

Methodology (I). Flow measures

Our sample includes the exact inflows and outflows. We normalise these flows dividing them by fund size (or number of investors)

□ But we also analyse the implicit flows:

$$IPMF_{it} = \frac{TNA_{it} - TNA_{i,t-1}(1+R_{it}) - MGTNA_{it}}{TNA_{i,t-1}}$$
$$IPIF_{it} = \frac{I_{it} - I_{i,t-1} - MGINV_{it}}{I_{i,t-1}}$$

Metodology (II). Performance measures Excess return (over the MSCI Spain Index) The alpha of the single factor model (CAPM) $R_{it} - R_{ft} = \alpha_i^1 + \beta_i^1 (R_{mt} - R_{ft}) + \varepsilon_{it}$ The alpha of 3-factor model (Fama-French, 1993) $R_{it} - R_{ft} = \alpha_i^3 + \beta_{iRMRF}^3 RMRF_t + \beta_{iSMB}^3 SMB_t + \beta_{iHML}^3 HML_t + \varepsilon_{it}$ The alpha of 4-factor model (Carhart, 1997) $R_{it} - R_{ft} = \alpha_i^4 + \beta_{iRMRF}^4 RMRF_t + \beta_{iSMB}^4 SMB_t + \beta_{iHML}^4 HML_t + \beta_{iPR1YR}^4 PR1YR_t + \varepsilon_{it}$

New money/investors vs old money/investors

- We follow the approach of Keswani and Stolin (2008) comparing the performance of new money portfolios and old money portfolios
- We can also analyse investor portfolios
- Our approach is based on monthly crosssectional comparison of:
 - TNA (investors) weighted portfolios \rightarrow Old M/I
 - Inflow-weighted portfolios \rightarrow New (In) M/I

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 - TNA (investors) weighted portfolios → Old M/I
 - Inflow-weighted portfolios → New (In) M/I
 - Outflow-weighted portfolios \rightarrow Out M/I

(I) 3-month and 12-month holding periods present significant negative performance

3-month holding period							
	ER	α1	α3	α4			
(1) EWP	-0.0045	-0.0017	-0.0101	-0.0086			
	(.017)	(.239)	(.000)	(.000)			

12-month holding period							
	ER	α1	α3	α4			
(1) EWP	-0.0200	-0.0102	-0.0441	-0.0426			
	(.000)	(.003)	(.000)	(.000)			

(II) Large funds present worse performance

	ER	α1	α3	α4
(1) EWP (3-months)	-0.0045	-0.0017	-0.0101	-0.0086
(2) TNA-weighted	-0.0068	-0.0034	-0.0116	-0.0090
	(.354)	(.427)	(.499)	(.865)
(5) Inv-weighted	-0.0083	-0.0060	-0.0142	-0.0126
	(.109)	(.033)	(.053)	(.056)
(1) EWP (12-months)	-0.0200	-0.0102	-0.0441	-0.0426
(2) TNA-weighted	-0.0287	-0.0186	-0.0510	-0.0488
	(.101)	(.111)	(.264)	(.325)
(5) Inv-weighted	-0.0344	-0.0276	-0.0601	-0.0588
	(.006)	(.001)	(.007)	(.007)

(III) Evidence of smart new (not out) money

	ER	α1	α3	α4
(2) TNA-weighted (3-months)	-0.0068	-0.0034	-0.0116	-0.0090
(3) Weighted by money in	-0.0010	0.0015	-0.0068	-0.0044
	(.056)	(.051)	(.066)	(.088)
(4) Weighted by money out	-0.0046	-0.0018	-0.0102	-0.0082
	(.416)	(.503)	(.561)	(.763)
(2) TNA-weighted (12-months)	-0.0287	-0.0186	-0.0510	-0.0488
(3) Weighted by money in	-0.0035	0.0050	-0.0310	-0.0283
	(.000)	(.000)	(.004)	(.003)
(4) Weighted by money out	-0.0175	-0.0082	-0.0426	-0.0412
	(.042)	(.060)	(.183)	(.238)

(IV) Evidence of smart new (not out) investors

	ER	α1	α3	α4
(5) Investor-weighted (3-months)	-0.0083	-0.0060	-0.0142	-0.0126
(6) Weighted by inv. in	0.0019	0.0037	-0.0052	-0.0009
	(.002)	(.000)	(.000)	(.000)
(7) Weighted by inv. out	-0.0055	-0.0019	-0.0106	-0.0073
	(.243)	(.048)	(.098)	(.012)
(5) Investor-weighted (12-months)	-0.0344	-0.0276	-0.0601	-0.0588
(6) Weighted by inv. in	0.0043	0.0125	-0.0241	-0.0197
	(.000)	(.000)	(.000)	(.000)
(7) Weighted by inv. out	-0.0201	-0.0094	-0.0437	-0.0404
	(.004)	(.000)	(.006)	(.002)

(V) New investors seem to be

smarter than new money...

	ER	α1	α 3	α4
(3-months)				
(3) Weighted by money in –(6) Weighted by investors in	-0.0029	-0.0022	-0.0016	-0.0036
	(.445)	(.404)	(.562)	(.199)
(12-months)				
(3) Weighted by money in –(6) Weighted by investors in	-0.0078	-0.0075	-0.0070	-0.0086
	(.352)	(.249)	(.323)	(.227)

Results. New (Away) M/I vs Old M/I

(V) New investors seem to be

smarter than new money...

	ER	α1	α 3	α 4			
(3-months)							
(3) Weighted by money in –(6) Weighted by investors in	-0.0029	-0.0022	-0.0016	-0.0036			
	(.445)	(.404)	(.562)	(.199)			
(12-months)							
(3) Weighted by money in –(6) Weighted by investors in	(3) Weighted by money in – (6) Weighted by investors in -0.0078 -0.0075 -0.0070 -0.008						
(.352) (.249) (.323) (.227)							
But we can't find significant evidence							

Positive flow portfolios vs negative flow portfolios

- We analyse the smartness of flows from a longshort strategy perspective
- This is the usual approach in financial literature, hence we also consider implicit flows to compare
- For each flow measure (implicit and exact money/investor flows), we rank funds:
 - with positive vs negative flows;
 - computing equally and flows weighted portfolios;
 - reporting performance differences (with sig.levels)

Results. Positive vs negative Implicit flows

<u>We are going to compute Excess Return</u> and α4 (12-month holding periods)

	Positive flow funds		Negativ fun	/e flow ds
	ER	α4	ER	α4
I Money				
I M weighted				
I Investors				
I I weighted				

Results. Positive vs negative Implicit flows

Positive flow portfolios present higher levels of Excess Return...

	Positive flow funds		Negativ fun	/e flow ds
	ER α4		ER	α4
I Money	-0.0093		-0.0267	
I M weighted	0.0088		-0.0217	
I Investors	-0.0055		-0.0308	
I I weighted	0.0168		-0.0249	

Results. Positive vs negative Implicit flows

... and also higuer levels of $\alpha 4$

	Positive flow funds		Negativ fun	/e flow ds
	ER	α4	ER	α4
I Money	-0.0093	-0.0306	-0.0267	-0.0506
I M weighted	0.0088	-0.0177	-0.0217	-0.0475
I Investors	-0.0055	-0.0257	-0.0308	-0.0528
I I weighted	0.0168	-0.0106	-0.0249	-0.0465

Implicit flows are smart

Results. Positive vs negative Implicit flows

<u>Important: observe that weighted</u> portfolios always present better results

	Positive flow funds		Negativ fun	/e flow ds
	ER	α4	ER	α4
I Money	-0.0093	-0.0306	-0.0267	-0.0506
I M weighted	0.0088	-0.0177	-0.0217	-0.0475
I Investors	-0.0055	-0.0257	-0.0308	-0.0528
I I weighted	0.0168	-0.0106	-0.0249	-0.0465

→ largest flows are invested in the best performers

Results. Positive vs negative Implicit flows

Important question: are these differences statistically significant?

Results. Positive vs negative Implicit flows

Important question: are these gaps statistically significant?

	Positive flow funds		Negative flow funds		Differences Pos – Neg	
	ER	α4	ER	α4	ER	α4
I Money	-0.0093	-0.0306	-0.0267	-0.0506	0.0174	0.0199
I M weighted	0.0088	-0.0177	-0.0217	-0.0475	0.0305	0.0298
I Investors	-0.0055	-0.0257	-0.0308	-0.0528	0.0252	0.0271
I I weighted	0.0168	-0.0106	-0.0249	-0.0465	0.0417	0.0359

→ We find significance at 1% level

Results. Positive vs negative Implicit flows: money vs investors

Can we find differences statistically significant?

		Differences Pos – Neg			
		ER	α4		
(1)	I Money	0.0174	0.0199		
(2)	I M weighted	0.0305	0.0298		
(3)	I Investors	0.0252	0.0271		
(4)	I I weighted	0.0417	0.0359		

Results. Positive vs negative Implicit flows: money vs investors

Can we find differences statistically significant?

		Differences Pos – Neg					
		ER	α4				
(1)	I Money	0.0174	0.0199		1-3	-0.0078	-0.0072
(2)	I M weighted	0.0305	0.0298			0.000	0.000
(3)	I Investors	0.0252	0.0271		2-4	-0.0112	-0.0061
(4)	I I weighted	0.0417	0.0359			0.074	0.244

In equally-weighted portfolios

Results. Positive vs negative *Exact flows*

An additional finding is related to the similar results that we can observe when considering exact flows.

Hence, we are providing evidence of the limited bias that prior studies have suffered when carrying out these analyses Results. Positive vs negative Exact flows

An additional finding is related to the similar results that we can observe when considering exact flows.

→ Hence, we are providing evidence of the limited bias that prior studies have suffered when carrying out these analyses but our study presents a limited bias

since it considers monthly information

Individual analyses

- Our study also presents another original approach of smart money: individual analyses
- While prior literature focuses on a global perspective, we consider both a time-series and a cross-sectional point of view
- The first analysis aims at detecting investors' timing abilities considering each fund separately
- On the other hand, the second approach is devoted to find possible selection abilities in each period of the sample

Individual analyses: Investors' timing abilities

This time-series approach tries to analyse if investors are able to choose the best moments to invest or divest from a fund:

$$P_{i,t+1} - \overline{P}_{t+1} = \alpha_t^1 + \beta_t^1 \left(F_{it} - \overline{F}_t \right) + \varepsilon_t$$

- For each fund, we calculate if prior excess flows anticipate subsequent excess performance
- Observe that flows are computed in relative terms to allow the comparison of all de funds of the category

Individual analyses: Investors' selection abilities

This analysis tries to shed additional light about the possible smartness of investors when selecting among all the available portfolios

$$P_{i,t+1} - \overline{P}_{t+1} = \alpha_t^2 + \beta_t^2 \left(F_{it} - \overline{F}_t \right) + \varepsilon_t$$

Again, in this cross-sectional analysis, we calculate if prior excess flows anticipate subsequent excess performance in each month

In both analyses, positive and significant betas would provide evidence of smart decisions

Results: No evidence of timing abilities *e.g.: 3-month holding period*

		Positive β^{1}		Negative eta ¹	
		Total	Sign.	Total	Sign.
Excess Return	Implied money flow	59	6	53	6
	Implied investor flow	48	6	64	5
	Money inflows	52	9	60	10
	Investor inflows	40	6	72	5
$lpha_4$	Implied money flow	55	8	57	9
	Implied investor flow	52	4	60	9
	Money inflows	51	13	61	10
	Investor inflows	47	5	65	8

Results: Some evidence of selection abilities *e.g.: 12-month holding period*

		Positive β^2		Negative β^{2}	
		Total	Sign.	Total	Sign.
Excess Return	Implied money flow	73	16	12	0
	Implied investor flow	71	24	14	0
	Money inflows	76	12	9	0
	Investor inflows	73	23	12	2
$lpha_4$	Implied money flow	70	17	15	0
	Implied investor flow	70	21	15	0
	Money inflows	72	11	13	1
	Investor inflows	75	20	10	1

Conclusions (I)

- The main aim of our study is focused on the determination of the possible investors' abilities to anticipate superior portfolio performance
- Our analyses present some relevant originalities:
 - \rightarrow Our dataset includes information of number of <u>investors</u> as well as the usual related to money
 - \rightarrow Our calculations are considered in <u>relative</u> terms
 - \rightarrow We have <u>exact</u> information of inflows and outflows
 - → We calculate *four* classes of performance
 - → We analyse Smart Money from an *individual* perspective

Conclusions (II) New flow performance

We provide general evidence of smart inflows

- This smartness is more marked in 12-month holding periods and for investor flows
- However, we fail to find statistical significance of superior abilities of investors vs. money
- Results obtained with outflows need further research

Conclusions (III) Long-short strategy

- We find that portfolios with positive flows obtain superior performance than those with negative flows
- These better results are statistically significant
- Moreover, largest flows are invested in the best performers
- These findings are significantly more marked when considering investor flows

Conclusions (IV) Individual perspective

- We propose two innovative approaches: a timing perspective for each fund and a selection method in each month
- The first approach does not provide evidence of timing abilities
- However, the second perspective shed more light about the underlying reasons of the Smart Money observed in the overall methods

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