SME investment best strategies
Outliers for assessing how to optimize performance.

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Abstract

Any research on strategies for reaching business excellence aims
at revealing the appropriate course of actions any executive should
consider. Thus, discussions take place on how effective a performance
measurement system can be estimated, or/and validated. Relevant
questions can be raised, like: can one find an adequate measure (i) on
the performance result due to whatever level of investment, and (ii) on
the timing of such investments? We argue that extreme value statis-
tics provide the answer. We demonstrate that the level and timing
of investments allow to be forecasting small and medium size enter-
prises (SME) performance, - at financial crisis times. The "invest-
ment level" is taken as the yearly total tangible asset (TTA). The
financial/economic performance indicators defining "growth" are the

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sales or total assets variations; "profitability" is defined from returns on investments or returns on sales. Companies on the Italian Stock Exchange STAR Market serve as example. It is found from the distributions extreme values that outlier companies (with positive performance) are those with the lowest but growing TTA. In contrast, the SME with low TTA, but which did not increase its TTA, before the crisis, became a "negative outlier". The outcome of these statistical findings should suggest strategies to SME board members.

Keywords: econophysics; outliers; STAR market; SME; strategies.

1 Introduction

The statistic and econophysic literature is filled with hundreds of papers on how to apply methods in order to measure, assess, discuss the reliability of financial data, (Amendola et al., 2006; Amendola et al., 2008; Boente et al., 2010; Clippe and Ausloos, 2012; Shang and Wang, 2013; Mir et al., 2014; Ausloos et al., 2016, 2017; Ceptureanu et al., 2017a, 2017b; Riccioni and Cerqueti, 2018; Shi et al., 2018) and develop strategies or suggest hints toward "better performance". Indeed, any research on strategies for reaching business excellence under an economic crisis aims at revealing the appropriate course of actions which any executive should consider, under the limitations and particular conditions that arise in an economic crisis environment (Afthonidis and Tsiotras, 2014). The first reaction is often to implement cost saving policies, to interrupt investment plans and proceed to business restructuring with cost cutting in mind. This may have some direct results, yet will not secure the future of the enterprise (Koksal and Ozgul, 2007). Others consider that the first step of the management should be to secure adequate resources, especially liquidity, - but again, history has shown that, very rarely during a recession, has this defensive strategy brought satisfactory results in the long run (Reeves and Deimler, 2009).

Nevertheless, justifying an investment can be frustratingly difficult to suggest when the payback is measured by loosely convincing projected long-term increases in sales, assets, and other profitability performance measures. Thus, discussions take place on how effective a performance measurement system can be estimated, validated, or credible (Vitale and Mavrinac, 1995), - a huge statistical set of criteria!

Thus, despite the episodic pervasiveness of recessions and their destructive impact on firms, and such mentioned considerations, a void exists in the management literature examining the intersection between recessions, strategy, and performance (Behrens et al., 2004; Latham and Braun, 2011). In fact, it can be asked whether there is an "initial condition" which in this non-linear set of plans and activities determines, and the more so allow to forecast, the future performance; in other words (Bourne and Neely, 2001) one has to wonder why measurement initiatives succeed and fail. The more so, one should avoid extreme losses (Vaz de Melo Mendes, 2006) and rather aim at huge gains based on some *ad hoc* strategy (Roberts, 2003; Reeves and Deimler, 2009).

In the econophysics framework, the connections between recession, strategic actions and business performance have been widely explored. Tan et al. (2014) propose a long-term memory study of returns and volatility of financial stocks, and argue that economic crisis shows a remarkable impact on emerging markets and it prevents a reliable stock returns prediction. On this line, Arshav and Rizvi (2015) highlights that markets have better efficiency in boom than in recession, with a special focus on Asian markets. Hua et al. (2016) deal with the cyclical patterns of financial markets and advance a new proposal for stocks evaluation which is able also to detect economic booms and recessions.

For statistical purposes and reasoning, can one find an adequate measure (i) on the performance result due to whatever level of investment, and (ii) on the timing of such investments? Is it simply obvious that a control of the investment strategy will lead to an improved (hopefully optimal) performance? To find some answers to these questions leads to the aims of this report.

The information and feedback from the measures should be used to challenge the assumptions and test the validity of whatever strategy (Eccles and Pyburn, 1992; Kaplan and Norton, 1996; Feurer and Chaharbaghi, 1995). In fact, authors have argued that they should be used for both purposes (Feurer and Chaharbaghi, 1995; Kaplan and Norton, 1996). Therefore, assessing the implementation of strategy and challenging the strategic assumptions are the two main subdivisions of the use of the performance measures. On performance measure suggesting investment policies, at a time of crisis, let us not go back too far recalling history; select a recent one "THE crisis", for immediately connecting thoughts to the above questions. Let us rather consider the practical side of the investigation: the use of statistics in performance measures, allowing for survival (Datta et al., 2016).

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Focusing on SME, an increase performance may depend on very appropriate investment strategies, more than for multinational firms. Innovation input and development are surely traditional set-ups, but others can arise from "more internally based" pertinent strategies imagined by the executive board.

Neely has much elaborated on measuring operations performance of SMEs, - alone or with coworkers (Neely, 1997; Neely and Austin, 2002; Kennerley and Neely, 2002; Neely and Al Najjar, 2006), distinguishing various points of view and evaluating relevant filters for analysis, even employee and/or customer satisfactions. For completeness, and in view of the specificity of the report, using Italy stock market as the case, let us point also (because of the specificity of the study, the Veneto Region in Italy and the East of England, UK) to Neely et al. (2001) on the impact of innovation on business performance.

The measures used for business performance efficiency are coherent with the expected innovation outcomes described in the Oslo Manual (2005) and are widely adopted in literature on innovation and performance. Several scholars provide evidence about association between innovation initiatives and financial performance (see e.g. the very recent contributions of Ferraro and Iovanella, 2017; Bartolacci et al., 2017; Ausloos et al., 2018; Ramanathan et al., 2018). The exclusive consideration of financial indicators, which may limit the analysis from one side, is a consequence of the choice to base the study on publicly available information that, on the other hand allow for completeness and integrity of data.

The Italian stock market has been considered in previous contributes whose common trait is the use of applied statistics, with a special focus on network analysis. D'Errico et al. (2004) investigated to what extent centrality within shareholding network is associated to stock volatility; Garlaschelli et al. (2005) model shareholding networks to shed lights on the laws ruling portfolio structures; they also provide a comprehensive description of investors and companies; finally Rotundo and D'Arcangelis (2010) study portfolio diversification as a possible strategy exerted by companies to gain relevance on the market through control over other companies. Thus, recalling previous works in the field of performance measurement incites to consider "extreme profitability" as another measure beside "huge growth" among the indicators of interest. Whence, we should obtain some statistical inference using extreme order statistics (Caroni and Karioti, 2004; Unnikrishnan, 2010; Gumedze and Chatora, 2014). This is developed in Sections

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2-3. In Section 4, with some summary, we offer some conclusive remarks and provide also suggestions for future research directions.

2   The framework

"Despite the episodic pervasiveness of recessions and their destructive impact on firms, a void exists in the management literature examining the intersection between recessions, strategy, and performance" wrote Latham and Braun (2011). Our paper seeks to address and bridge this research gap, finding a statistical relationship between marketing strategies and firms performance near an expectedly forthcoming economic crisis. Such a focussed aim is also raised by Koksal and Ozgul (2007).

It seems that our common agreement on the most basic or pertinent questions is about the levels of investments that can be used, - and in what timing order. Obviously, one needs to rely on a preliminary acceptable measure of the so called "investment efficiency." Thereafter, the questions appear to be two fold, - with respect to the quantitative aspects: (1) Should one (later) measure the efficiency in terms of the lowest investment, or (2) *a contrario*, is a high amount of investment necessarily for a better performance? In fact, it can be hard to decide what consecutive investments (one "low" followed by one "high", or the other way around, or even with some longer cycling) is responsible for a subsequent efficiency.

The main point (H1) to be clarified pertains of course to the (statistics) definition of the "business performance efficiency" measure. It is here considered that only a few aspects seem relevant. In view of the pertinent literature, we select four variables, or financial/economic indicators, for representing business performance: two of them for "growth", which can be expressed through (i) sales variations (DS) and (ii) total assets variations (DA), and also two for "profitability", through (iii) returns on investments (ROI) and (iv) returns on sales (ROS).

Next, (H2) one can admit that a certain time span has to be used for obtaining a reliable measure. These indicators will be measured here from publicly available results (in 2008, 2009, and 2010), AFTER the crisis, averaged over such a 3 year time interval: the notation will be for example $< DS >_3$ for the sales (S) variations, averaged over 3 years: [2008-2010]. The variable of interest measuring some level of investments is taken to be the firm total tangible assets (TTA). The data of interest BEFORE the ("un-

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known" or incoming) crisis is chosen to cover 2006 and 2007. It will be noted as TTA06 or TTA07; moreover, its average is noted $< TTA >_2$.

2.1 Statistical analysis methodology

After having performed the 3 year averaging for (i)-(iv), the methodology goes as follows: each (i)-(iv) average values are used as the numerator of the "performance efficiency" ratio in which the denominator is either the lowest TTA (TTAm) or the highest TTA (TTAM), value in either 2006 or 2007. Thereafter, the (i)-(iv) averages of the firms are also compared with respect to the TTA average trough their ratio for which the denominator is $< TTA >_2$, equal to (1/2)(TTAm+TTAM), of course. This leads to 12 indicators. The (62, at that time) SME on the STAR Market Segment of the Italian Stock Exchange are considered to span various types of SME and a convenient sample for examining statistical characteristics leading to conclusions on performance efficiency. N.B. The STAR (Segment for High Requirement Shares) market
$http : //www : borsaitaliana : it/azioni/mercati/star/home-star/segmento-star : en.htm$ includes companies capitalized from 40 million to 100 million Euros; see: $http : //www : borsaitaliana.it/homepage/homepage.htm$ within the Milano electronic share market (Mercato Telematico Azionario: MTA): $http : //www : borsaitaliana : it/azioni/mercati/mta/home/mta-mercato-telematico-azionario : en.htm$

2.2 Stressing the usefulness of extreme values

A very fundamental point is next emphasized: it should be easily understood and accepted that the statistical outliers are the companies giving a better view of the success or failure of their previous investment strategy. The outliers overperform or underperform. That is what is usually to be avoided or searched for, whence to be attracting the discussion: means are often considered. However such values, whence firms, for which the final outcome occurs "near the average" are in fact "strategically uninteresting", - because merely falling within statistical error bars; thus, they should not be considered to be relevant for our purposes. Therefore, the outliers are next extracted, shone upon, and discussed for emphasizing the interesting features allowing recommendations.

Variable	Min.	Max.	Sum	Mean (μ)	StDev (σ)	Skewness	Kurtosis
$TTAm$	42.000	$4.829\ 10^5$	$2.600\ 10^6$	41931	89262	3.4022	11.948
$TTAM$	131.00	$5.321\ 10^5$	$2.893\ 10^6$	46662	96049	3.3905	12.156
$<TTA>_2$	86.5	$5.075\ 10^5$	$2.746\ 10^6$	44297	92600	3.3967	12.062
$<DS>_3$	-0.1924	1.1767	4.9303	0.0795	0.198	3.1414	14.013
$<DA>_3$	-0.1436	1.9818	7.8786	0.1271	0.330	3.8060	16.885
$<ROI>_3$	-0.0768	0.3457	3.0115	0.0486	0.067	1.5342	5.1206
$<ROS>_3$	-0.6609	0.2445	2.5316	0.0408	0.116	-3.505	20.046

Table 1: Summary of (rounded) statistical characteristics for the time average distributions of the growth and profitability indicators for the 62 STAR companies, and of their $<TTA>_2$, in the center of the table, in per cents and in 10^6 Euros, respectively; the skewness and kurtosis are dimensionless scalars.

This reasoning is in line with the statistical literature which includes work on exploring possible trends in damages resulting from extreme events, like earthquakes (Pisarenko and Sornette, 2003; Sornette and Werner, 2011; Ficcadenti and Cerqueti, 2017) or floods (Akinsete et al., 2008), and survival analysis (Datta et al., 2016).

3 Results analysis

The raw data main statistical characteristics are given in Table 1. Observe that since there is a negative minimum for each (i)-(iv), some board strategies were rather failures. Nevertheless, the mean is always positive. The distributions are quite extended, as indicated by the (easily estimated from the data in the table) so called coefficient of variation σ/μ values. The kurtosis is always positive and large, indicating lesser chances of extreme negative outcomes; the skewness is positive, indicating a long positive tail (many small losses and a few extreme gains), - except for $<ROS>_3$ which has an unexpected negative skewness, thereby indicating a long lower range tail (many small gains and several extreme losses). To have a paramount view of the data, see also the histograms reported in Figures 1-5.

Variable	Min.	Max.	Sum	Skewness	Kurtosis
$< DS >_3 \,/\, TTAm$	-0.01482	0.4795	1.1602	5.1151	24.804
$< DA >_3 /TTAm$	-0.11547	0.5089	0.4468	6.6860	48.151
$< ROI >_3 \,/TTAm$	-0.01313	0.1573	0.4007	4.8640	23.330
$< ROS >_3 \,/TTAm$	-0.24661	0.1623	0.1533	-2.2285	23.790
$< DS >_3 \,/TTAM$	$-8.91\ 10^{-3}$	0.3962	0.6726	6.3815	41.721
$< DA >_3 \,/TTAM$	-0.03702	0.4524	0.4827	7.3938	54.155
$< ROI >_3 \,/TTAM$	$-6.53\ 10^{-3}$	0.0733	0.1894	4.8538	24.937
$< ROS >_3 \,/TTAM$	-0.12260	0.0573	0.0748	-3.5113	27.249
$< DS >_3 /< TTA >_2$	$-9.70\ 10^{-3}$	0.4195	0.8094	5.7396	33.346
$< DA >_3 /< TTA >_2$	-0.05607	0.4790	0.4882	7.2609	52.930
$< ROI >_3 \,/< TTA >_2$	$-8.72\ 10^{-3}$	0.1000	0.2479	4.9740	25.738
$< ROS >_3 /< TTA >_2$	-0.16380	0.0788	0.0900	-3.3352	26.988

Table 2: Summary of (rounded) statistical characteristics for the 12 statistical indicators distributions of the growth and profitability measures for the 62 STAR companies; in the center of the table, data is given in per cents; the skewness and kurtosis are dimensionless scalars.

Indicator	Mean (μ)	St.Dev. (σ)	$\mu - 2\sigma$	$\mu + 2\sigma$
$< DS >_3 \,/\, TTAm$	$1.8713\ 10^{-2}$	0.08278	-0.14684	0.18427
$< DA >_3 /TTAm$	$7.2064\ 10^{-3}$	0.06747	-0.12774	0.14215
$< ROI >_3 \,/TTAm$	$6.4631\ 10^{-3}$	0.02612	-0.04577	0.05869
$< ROS >_3 \,/TTAm$	$2.4721\ 10^{-3}$	0.04138	-0.08029	0.08524
$< DS >_3 \,/TTAM$	$1.0849\ 10^{-2}$	0.05379	-0.09673	0.11843
$< DA >_3 \,/TTAM$	$7.7854\ 10^{-3}$	0.05810	-0.10841	0.12398
$< ROI >_3 \,/TTAM$	$3.0546\ 10^{-3}$	0.01127	-0.01949	0.02560
$< ROS >_3 \,/TTAM$	$1.2058\ 10^{-3}$	0.01933	-0.03746	0.03987
$< DS >_3 /< TTA >_2$	$1.3055\ 10^{-2}$	0.06071	-0.10836	0.13447
$< DA >_3 /< TTA >_2$	$7.8741\ 10^{-3}$	0.06190	-0.11593	0.13168
$< ROI >_3 \,/< TTA >_2$	$3.9985\ 10^{-3}$	0.01540	-0.02681	0.03481
$< ROS >_3 /< TTA >_2$	$1.4520\ 10^{-3}$	0.02597	-0.05049	0.05339

Table 3: Indicators confidence interval (rounded) limits (in per cents).

	Company Name			
efficiency	(11)	(13) Cairo	(58)	(45)
Indicator ratio	Buongiorno	Communication	Ternienergia	Mondo TV
$< DS >_3$ / $TTAm$	0.4795	(0.0186)	0.4457	(0.0769)
$< DA >_3/TTAm$	(-0.1155)	(-0.0217)	0.5089	(-0.0536)
$< ROI >_3$ /$TTAm$	0.1277	0.1573	(0.0345)	(-0.0130)
$< ROS >_3$ /$TTAm$	0.1623	0.1228	(0.0436)	-0.2466
$< DS >_3$ /$TTAM$	0.1537	(0.0087)	0.3962	(0.0382)
$< DA >_3$ /$TTAM$	(-0.0370)	(-0.0101)	0.4524	(-0.0266)
$< ROI >_3$ /$TTAM$	0.0409	0.0733	0.0306	(-0.0065)
$< ROS >_3$ /$TTAM$	0.0520	0.0573	(0.0388)	(-0.1226)
$< DS >_3/< TTA >_2$	0.2328	(0.0118)	0.4195	(0.0511)
$< DA >_3/< TTA >_2$	(-0.0561)	(-0.0138)	0.4790	(-0.0356)
$< ROI >_3$ /$< TTA >_2$	0.0620	0.1000	(0.0324)	(-0.0872)
$< ROS >_3/< TTA >_2$	0.0788	0.0781	(0.0410)	(-0.1638)

Table 4: Main positive and negative outliers of the growth variations and profitability efficiency indicators for the 62 STAR companies in per cent, i.e. those falling outside the interval $]\mu - 2\sigma, \mu + 2\sigma[$ corresponding to each ratio distribution. The data in parentheses correspond to those companies which are not truly outliers in a statistical sense for the index of interest, - but almost, like the inefficient Mondo TV.
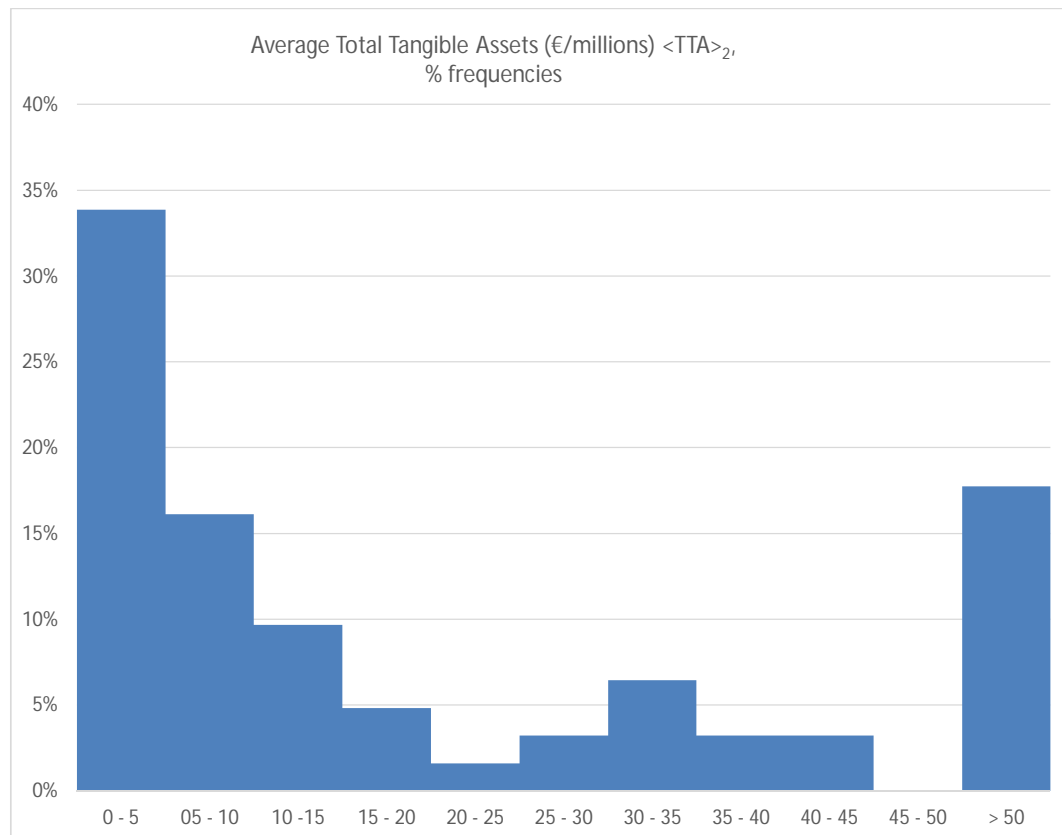
Figure 1: Histogram of the empirical distribution of $< TTA >_2$ for the 62 companies listed on the STAR market.
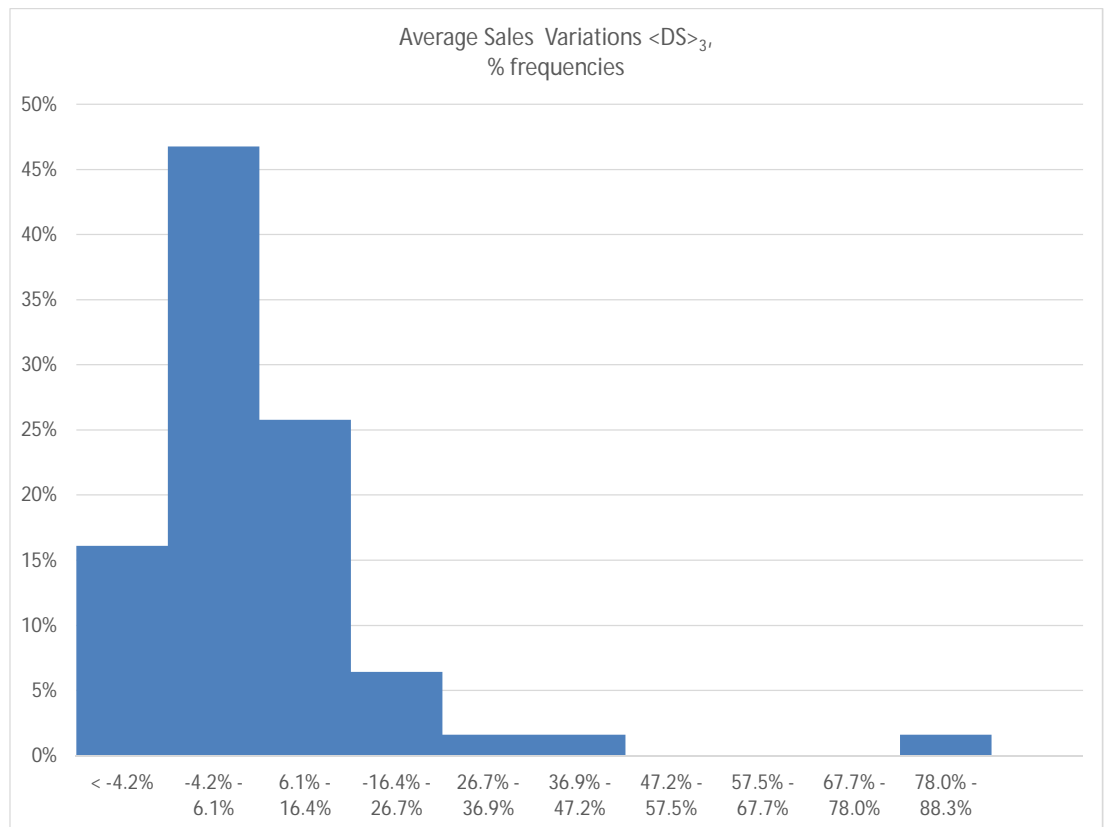
Figure 2: Histogram of the empirical distribution of $< DS >_3$ for the 62 companies listed on the STAR market.

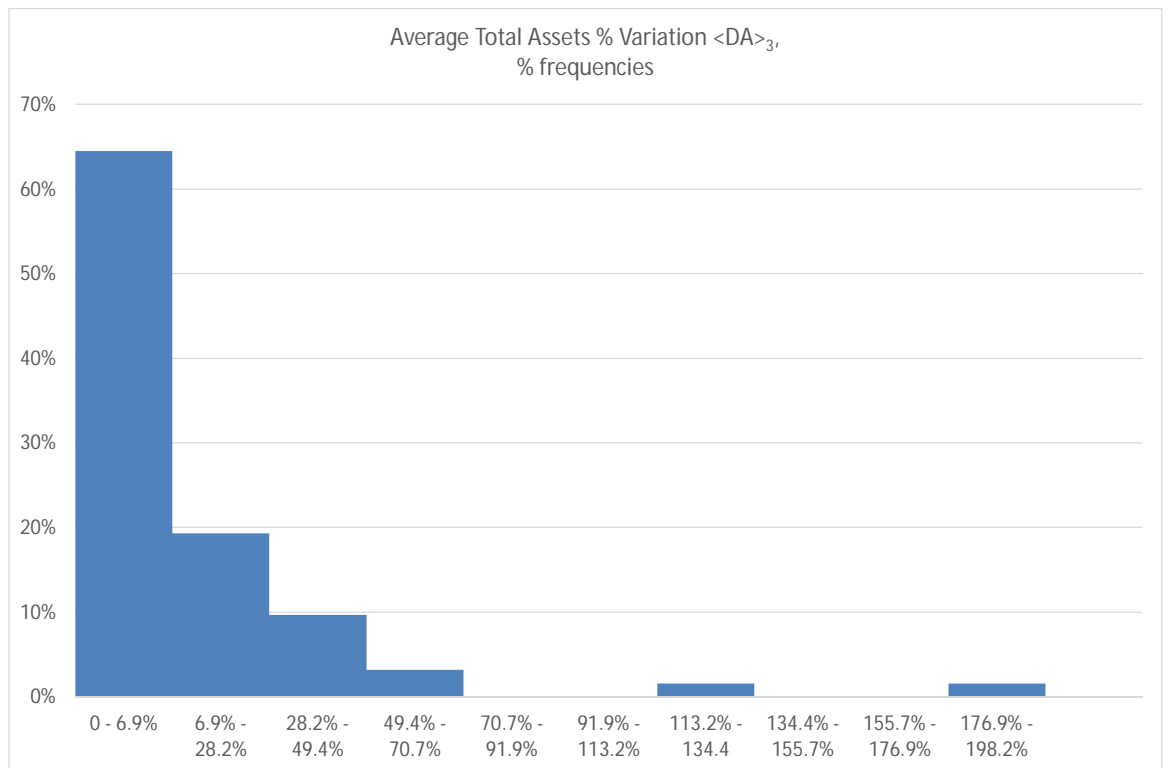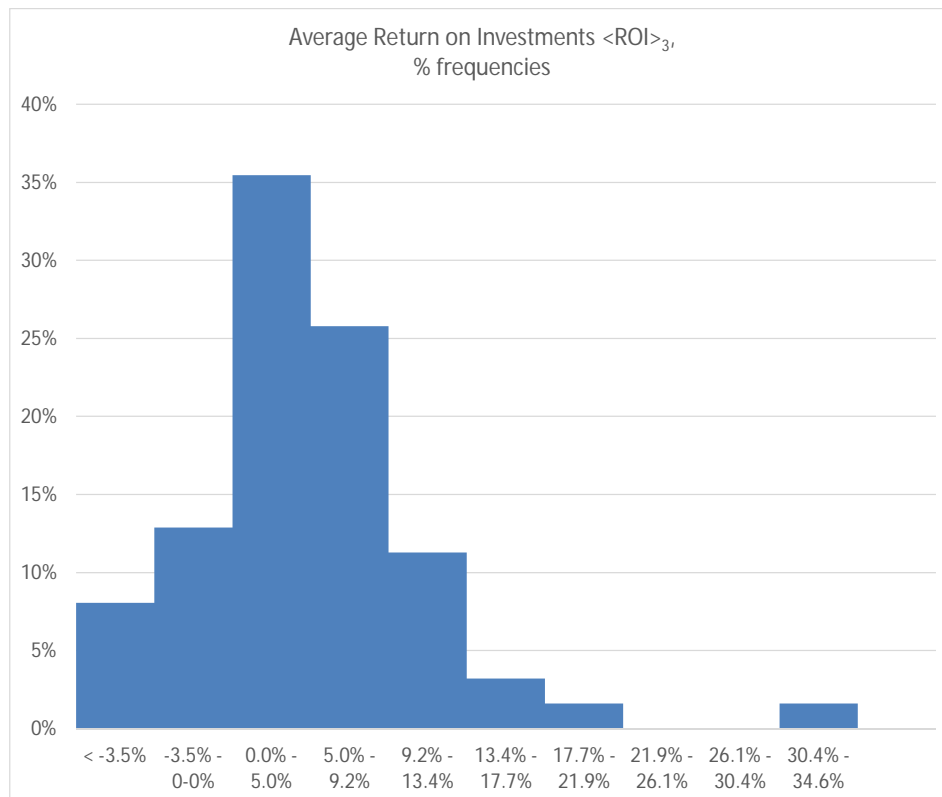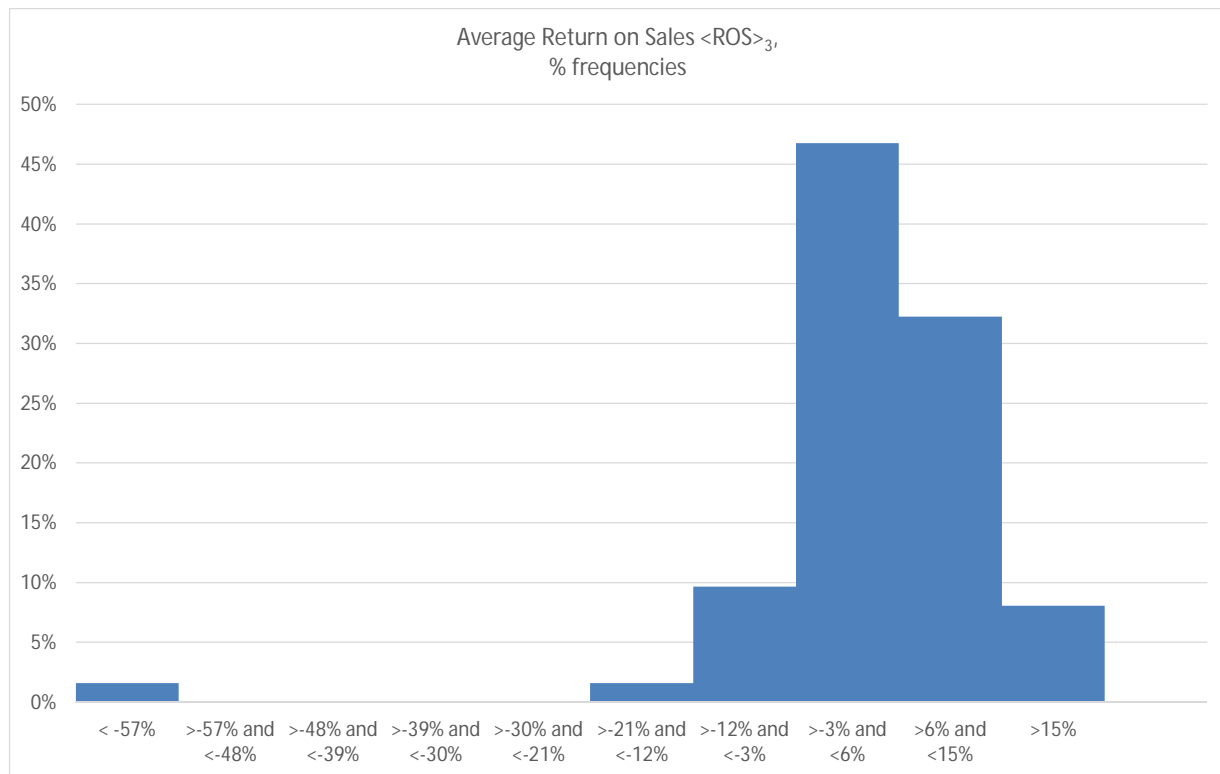Figure 3: Histogram of the empirical distribution of $< DA >_3$ for the 62 companies listed on the STAR market.

Figure 4: Histogram of the empirical distribution of $< ROI >_3$ for the 62 companies listed on the STAR market.

Figure 5: Histogram of the empirical distribution of $< ROS >_3$ for the 62 companies listed on the STAR market.
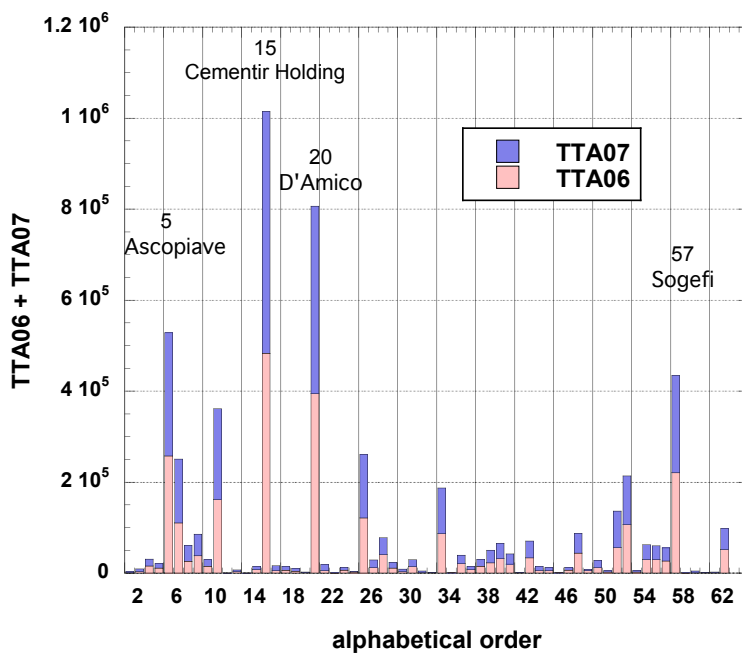
Figure 6: Stacked $TTA06$ and $TTA07$ of the 62 companies listed on the STAR market. x-axis is the alphabetical index for the 62 SMEs.'-. The four largest TTA firms are indicated.
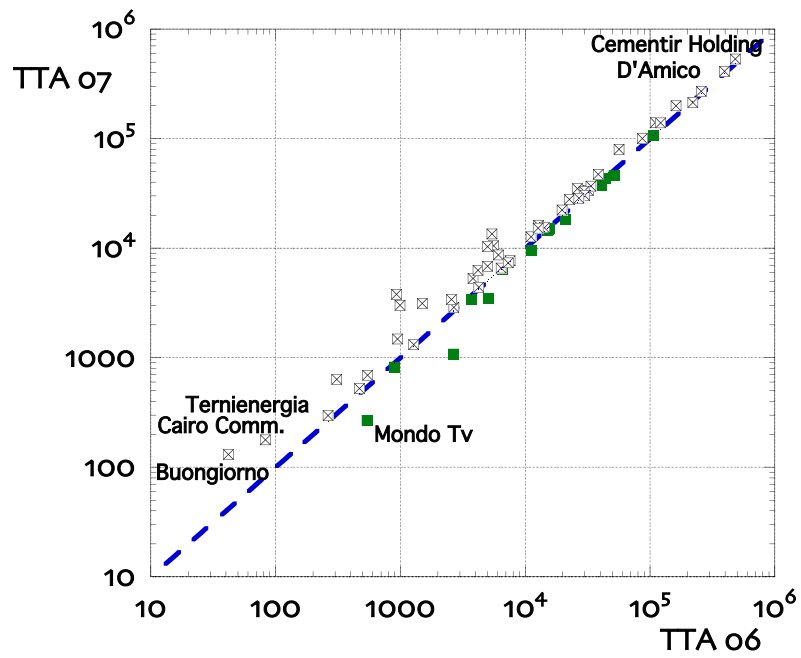
Figure 7: Relation TTA07 *vs.* TTA06 (in EUR) for the 62 companies listed on the STAR market, distinguishing between those with increased (open square with cross) or decreased (full square) TTA. The name of a few "interesting", thus "extreme", companies is pointed out.

Next consider whether some "cause" suggests why there are such outliers: the bar plot for the (stacked) TTA variables, in 2006 and 2007, is displayed in Figure 6. The major companies are pointed out. The (up or down) order of investments can be better observed on Figure 7 that, in 45 cases, there was an increase in TTA , i.e., TTA06 < TTA07 , and (of course) 17 cases are such that there was a decrease in TTA : TTA06 > TTA07 . It seems relevant to distinguish between these two categories in the discussion of features, - see below. The TTA magnitudes and the two types of investment classes can be distinguished in Figure 7. The statistical variations are not large, but not negligible.

The (rounded) statistical characteristics for the 12 statistical indicators distributions, hereby considered as a measure of the growth and profitability for the 62 STAR companies; are given in Table 2. N.B. data in the center of the table is given in per cents; the skewness and kurtosis are dimensionless scalars. The standard confidence intervals limits are readily deduced and reported in Table 3.

The performance efficiency ratios of the 62 companies are not given, for space savings, but those of the outliers, i.e. when the SME having efficiency values fall outside the relevant $]\mu - 2\sigma, \ \mu + 2\sigma[$ interval are listed in Table 4. There are 3 SMEs which are, rather systematically, positive outliers: (58) Ternienergia, (11) Buongiorno, (13) Cairo Communications, and 1 SME which is systematically "negative outlier": (45) Mondo TV. For completeness, we also display, in Table 4, the corresponding values for such companies, even when they are not true outliers in a statistical sense. It is found that all of these are usually close to the end of the statistical confidence interval; see Table 3. This is particularly the case of Mondo TV, for which all efficiency ratios, except for those involving $< DS >_3$, are negative. However, we repeat: such values almost fall within the statistical error bars deduced for the whole 62 firm set.

Interestingly, (11) Buongiorno appears most of the times in the top brackets, but appears at the bottom (the worst) for ratios involving $< DA >_3$. Another interesting finding concerns Buongiorno which appears as "almost a negative outlier" in three efficiency ratios; see Table 3 and 4. On the other hand, (58) Ternienergia and (13) Cairo Communications have very dissimilar performance efficiency behaviors: the former performing better for "growth", the latter performing better for "profitability". Due to the presence of such outliers, it is of course ridiculous to attempt a regression-like study. The resulting coefficients are all pointing to a valid null hypothesis. Nevertheless, it

should occur to the reader that those 4 companies are those with the lowest TTA; see Figure 7. Moreover, Mondo TV is the only one among the outliers which has a TTA06 higher than its TTA07, - this SME had about a 50% decrease in investment before the crisis. In contrast, Ternienergia, Buongiorno, and Cairo Communications have relatively the highest increases in TTA .

Results of correlations can be illustrated through figures, on which the highest TTA firms are more easily distinguished. However, in view of the above and Table 2, it should occur to the reader that such companies had not a well performing strategy. Indeed, a few of these "not systematically outlier companies" have a mixture of positive (or negative) small efficiency ratio values. One should observe that

- Figure 8 displays the relationship between $< DA >_3$ and $< TTA >_2$; the largest $< DA >_3$ effect occurs for Esprinet and Ternienergia, both with a low $< TTA >_2$. A small negative $< DA >_3$ for D'Amico which has a large $< TTA >_2$ is observed, in contrast to Cementir Holding and Ascopiave which have a large $< TTA >_2$ also, but with a slightly positive $< DA >_3$;

- Figure 9 displays the relationship between $< DS >_3$ and $< TTA >_2$: a large $< DS >_3$ effect occurs for Ternienergia (recall that it has a low $< TTA >_2$, as already emphasized); a negative $< DS >_3$ effect occurs for D'Amico and Cementir Holding;

- Figure 10 displays the relationship between $< ROI >_3$ and $< TTA >_2$: a weak $< ROI >_3$ effect is found for Cementir Holding and Ascopiave; a negative but much larger occurs for D'Amico; in contrast, a large $< ROI >_3$ occurs for Tesmec, while the negatively largest $< ROI >_3$ is for Eems, - both firms with rather low $< TTA >_2$;

- Figure 11 displays the relationship between $< ROS >_3$ and $< TTA >_2$; a moderate $< ROS >_3$ positive effect occurs for Sogefi, Ascopiave, D'Amico and Cementir Holding, the four largest TTA companies; a large negative $< ROS >_3$ effect occurs for Mondo TV; on the opposite side, the best $< ROS >_3$ positive effect is for Falck Renewables, Zignago Vetro, and Nice.
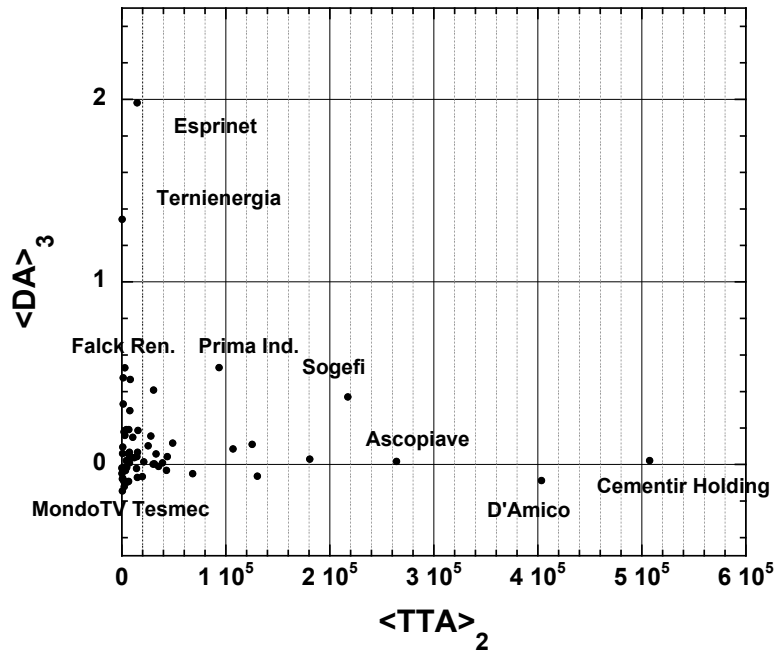
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Figure 8: Relation $< DA >_3$ *vs.* $< TTA >_2$ for the 62 companies listed on the STAR market: observe a large $< DA >_3$ effect for Esprinet and Ternienergia without much $< TTA >_2$; a small negative $< DA >_3$ with large $< TTA >_2$ for D'Amico and slightly positive $< DA >_3$ with large $< TTA >_2$ for Cementir Holding and Ascoclave.

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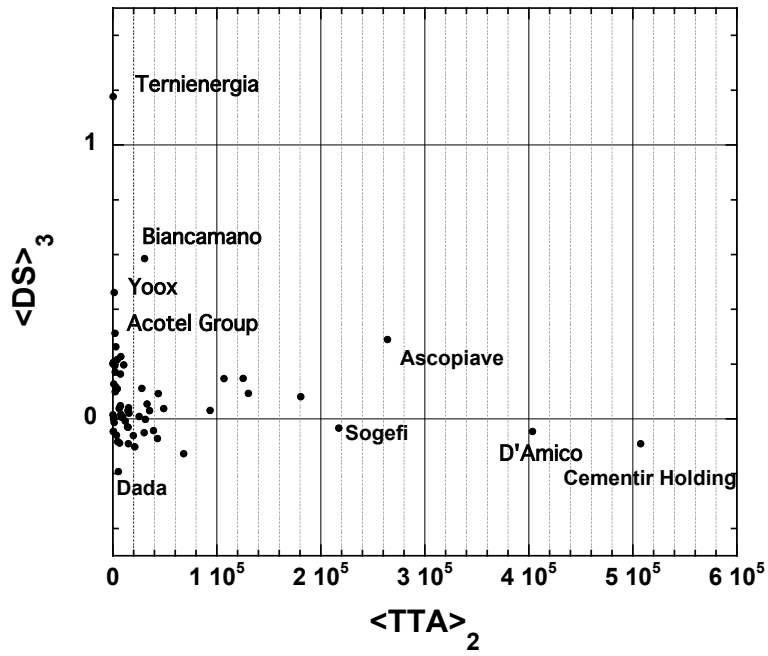
Figure 9: Relation $< DS >_3$ $vs.$ $< TTA >_2$ for the 62 companies listed on the STAR market: observe a large $< DS >_3$ effect for Ternienergia without much $< TTA >_2$; a negative $< DS >_3$ with large $< TTA >_2$ for D'Amico and Cementir Holding .
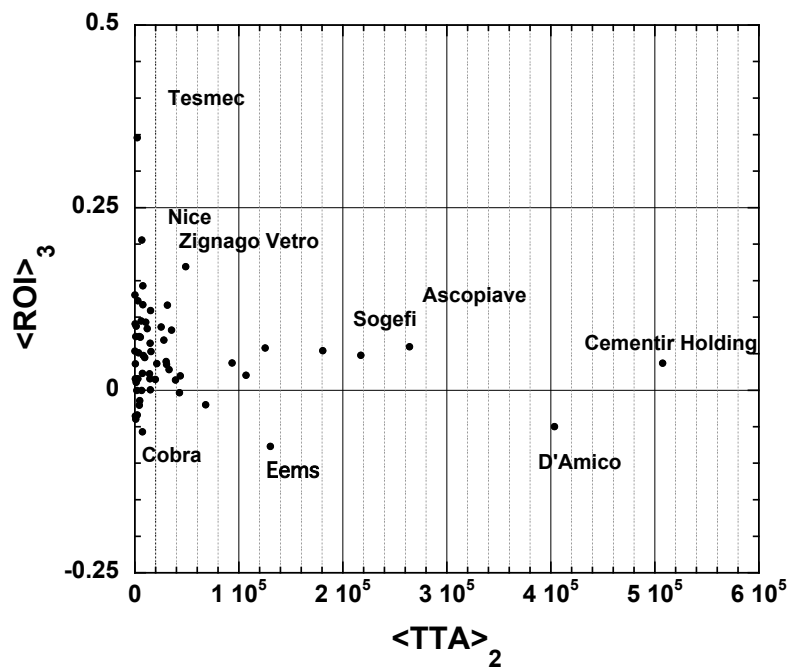
Figure 10: Relation $< ROI >_3$ $vs.$ $< TTA >_2$ for the 62 companies listed on the STAR market: observe a weak$< ROI >_3$ effect with large $< TTA >_2$ for Cementir Holding and Ascopiave; a much larger but negative for D'Amico; a large $< ROI >_3$ with low$< TTA >_2$ for Tesmec; the negative largest $< ROI >_3$ for Eems.
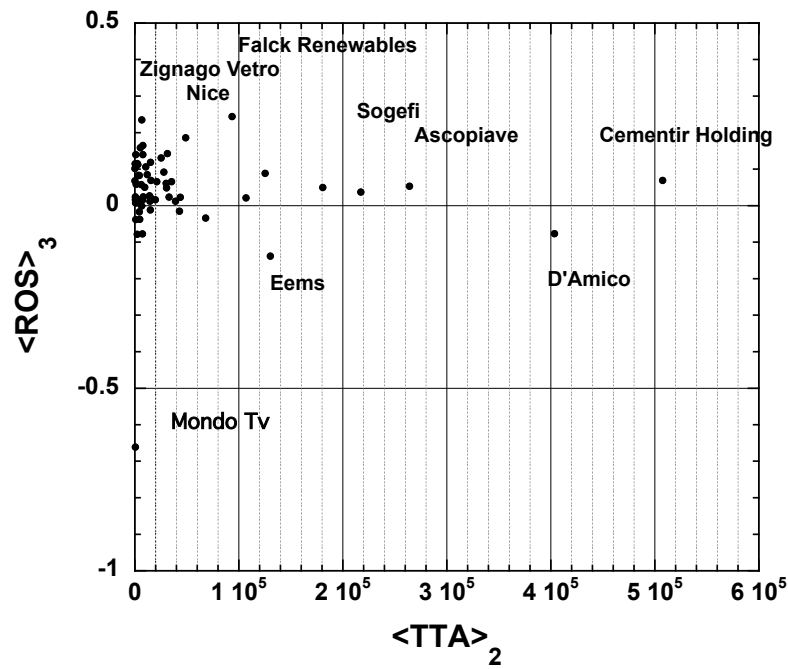
Figure 11: Relation $< ROS >_3$ $vs.$ $< TTA >_2$ for the 62 companies listed on the STAR market: see a moderate $< ROS >_3$ effect for Sogefi, Ascopiave, D'Amico and Cementir Holding; a large negative $< ROS >_3$ effect for Mondo TV; the best $< ROS >_3$ effect for Falck Renewables, Zignago Vetro, and Nice.

4 Conclusion

In this paper, we have proposed arguments in favor of extreme values to shine light on performance checking. The fact that a comprehensive set of outlier properties can be derived for measured "anomalous performance ratios" is a considerable attraction. Recall that these include two aims presently envisaged. Finding a convenient measure of investment performance, - whatever the investment, and from such a measure observe at the time of crisis what positive or negative effect has an investment "cause". The key timing separating the cause and its effect is the financial crisis. Notice that the study allows three considerations from extreme value analysis: not only the investment evolution; up or down, low or high, but also through their average, serving as a control kind of test. It should be obvious that the best performance should be better appreciated when (unexpectedly?) the investment is low.

For further introducing the following discussion, let us briefly define, in Table 5, the type of companies mentioned here above. Observe that they cover various sectors. This allows us to consider that the 62 STAR market companies represent an interesting and valuable set of SMEs for our investigation. Let it be observed that the positive outliers belong to different activities: Ternienergia: Utilities; Buongiorno: Technology; Cairo Communications: Media, while the negative outlier Mondo TV is also a Media actor. Therefore, a "SME segment independent universal rule" is found: all those 4 companies have the lowest TTA of the STAR market; recall Fig. 7.

Nevertheless, there are differences: Ternienergia and Cairo Communications have very dissimilar performance efficiency behaviors, the former performing better for "growth", the latter for "profitability". Since Ternienergia, Buongiorno, and Cairo Communications have a high increase in TTA, one might recommend such a strategy. In fact, Mondo TV did not increase its TTA, pointing to a deficient strategy, - again pointing that the timing of "investment" seems relevant; not the average value.

Conclusions follow, expecting that SME board members understand statistical facts about successful and unsuccessful strategies.

Specifically, one of the targets of the present paper has been to demonstrate (and discuss) the effect of a cause (found to be the assets) on the statistical mean of performance variables (either growth or profitability).

In fine, observe that this paper has been arguing that it is "truly interesting" to look at the extremes in distribution tails, indeed (Caroni and

Karioti, 2004; Unnikrishnan, 2010; Gumedze and Chatora, 2014). The fact that a comprehensive set of theoretical properties can be derived from extreme values is a considerable attraction. Even more important, from the practical point of view, is that our results provide more evidence on deducing some new principle, - here in order to optimize strategies within a forecasting perspective (Caldeira et al., 2016; Kapetanios et al., 2016), but surely in other cases as well. The statistical analysis in this paper points that the best resistance to crisis is found to occur for the firms with the initially lowest assets. An extremely important point resides in the timing of investments. An increase leads to a much better performance than a decrease starting from a higher level. We conclude that applied statistics is relevant in the context of performance strategy.

i:	Name	"Super sector"
1	Acotel Group	Telecommunications
5	Ascopiave	Utilities
7	Biancamano	Industrial goods & Services
11	Buongiorno	Technology
13	Cairo Communication	Media
15	Cementir Holding	Constructions & Materials
17	Cobra	Industrial goods & Services
18	Dada	Industrial goods & Services
20	D'Amico	Industrial goods & Services
25	Eems	Technology
30	Esprinet	Technology
31	Eurotech	Technology
33	Falck Renewables	Utilities
45	Mondo TV	Media
46	Nice	Industrial goods & Materials
50	Prima Industrie	Industrial goods & Materials
57	Sogefi	Automobiles & Parts
58	Ternienergia	Utilities
59	Tesmec	Industrial goods & Services
61	Yoox	Retailer
62	Zignago Vetro	Industrial goods & Services

Table 5: A few STAR company names which are mentioned in the text, or in figures, in alphabetical order (index i), and their business type.

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