Fair Value Hierarchy in Financial Instruments Disclosure. Is Transparency Well Assessed For Investors? Evidence from Banking Industry

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Abstract

The debate on fair value accounting is still open although last 20 years have been spent in searching for solution by academician, practitioners and institutions. After a long and continuous discussion both on the basic concepts and the information level contained in fair value measurements and on the different solutions that is possible to adopt in mark to market measurements, IASB and FASB have recently issued new standards on fair value measurements applying some principles not only to financial instruments but also to property and other investments. To verify if the solutions adopted in these Standards really improve the disclosure level and the "usefulness of data for investors", this paper tries to analyze the actual level of transparency and the "usefulness" of the "fair value hierarchy" (that for some aspects synthesized the Boards way of thinking referring to fair value) that has been already introduced for financial instruments by IFRS 7, Financial Instruments: Diclosure.

The paper presents results of an empirical investigation on a sample of domestic and foreign listed banks who adopted fair value hierarchy in line with SFAS 157 and IFRS 7 recommendations. Research questions can be summarized as follows: (i) does fair value hierarchy improve transparency in financial instruments evaluation in banks annual report or it can be considered as a tool for earnings management?

Keywords: Fair Value Hierarchy, Fair Value Measurement, IFRS7, Disclosure requirements

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1. Introduction

During this period of global markets, multinational corporations are demanding financial accounting standards with enhanced uniformity. In an effort to achieve this objective, the International Accounting Standards Board (IASB) and the Financial Accounting Standards Board (FASB) have been working together on the Convergence Project, aiming to develop accounting standards that closely correlate with international financial reporting standards. In September 2006 and February 2007, the FASB issued two key fair value accounting (FVA) standards which focused on providing guidelines for fair value measurement (through a classification hierarchy), expanding disclosure requirements, and also allowing business entities to increase FVA's application. In 2006 IASB issued IFRS 7, *Financial Instruments: Disclosures* asking to firms to provide market risk disclosures. The standard came into effect in January 2007. Last modifications to the standard have been applied on July 2008. It addresses the demand for risk information that allows investors and other classes of stakeholders to assess the firm's future economic performance adopting FVA. However, the recent financial crisis has placed increased scrutiny on estimates derived under FVA. Fair value measurement not limited to financial instruments is provided by IFRS 13, issued on 12 may 2011. However, the principle has not been endorsed by EU, even if Efrag provided endorsement advice on January 2012.

Two competing world views are identified as underlying the debate: a Fair Value View, implicit in the IASB's public pronouncements, and an Alternative View implicit in publicly expressed criticisms of the IASB's pronouncements. Whittington (2008) concluded that, in a realistic market setting, the search for a universal measurement method may be fruitless and a more appropriate approach to the measurement problem might be to define a clear measurement objective and to select the measurement method that best meets that objective in the particular circumstances that exist in relation to each item in the accounts.

There is a strong pressure from investors to report accounting items using fair value concept upon economic boom. The financial crisis period may raise an issue of revival of conservative concepts in financial reporting, e.g. historical costs measurement and application of prudence principle.

Considering financial instruments evaluation, since many years there has been an open discussion between those who consider that the mixed model¹ should definitely evolve into a full fair value model in which all financial instruments are measured at fair value with changes recognized in profit and loss account and those who consider that the mixed model is the optimal model for financial instruments accounting.

It seems that the full fair value model has clear advantages that are easy to demonstrate, but it is also true that there are sufficient disadvantages that advise against its implementation, at least for the moment. To limit the disadvantages joined to the difficulties to obtain "correct" estimates of fair values evaluated with models, the Boards decided to establish a three-level hierarchy that distinguishes (1) readily observable measurement inputs from (2) less readily observable measurement inputs and (3) unobservable measurement inputs.

No uniform framework is available to assure consistent fair market valuation and transparency for investor decision-making. Conceptual solution of valuation issues need not to come out from current economic situation and it is impossible to change this concept every time when economic conditions tend to change. Unsystematically changes of valuation concepts may conduce to instability of economic system.

2. Literature review

Main literature of last year considers different topics on fair value. For the purpose of this study we can consider investigations related to the value relevance of Fair Value Option (FVO) with particular regard to the effect of using FV adoption on earnings management practices end on quality of disclosure for investors.

¹ A mixed model is an accounting model that considers a mix of fair value and hystorical cost – depending from assets and liabilities - but also a model in which some modifications of fair value are in OCI instead of in Profit and Loss.

Fair Value Hierarchy, introduced by FAS No. 157 and transposed by IASB in IFRS 7, prioritizes the source of information used in fair value measurements into three levels: (1) Level 1 (observable inputs from quoted prices in active markets), (2) Level 2 (indirectly observable inputs from quoted prices of comparable items in active markets, identical items in inactive markets, or other market-related information), and (3) Level 3 (unobservable, firm-generated inputs). Considering the value relevance of fair value accounting, Song et al. (2008), using quarterly reports of banking firms in 2008, found that the value relevance of Level 1 and Level 2 fair values is greater than the value relevance of Level 3 fair values. This evidence produced a growing literature focused on the effect that the adoption of fair value measurements, with particular regard to financial instruments measurement considering financial crisis of last years, on reported earnings and on dividend policy adopted by listed companies and banks.

Starting from the question if fair value can helps earning manipulation, the study of Benston (2008) answer with a positive response, underlining that that implementation of SFAS 157 - who specifies the *fair value* as an *exit values* - is likely to be costly to investors and independent public accountants.

Using an international sample of 222 banks from 41 countries, the study of Fiechter (2011) examines whether the use of fair value option affects earnings volatility. Prior empirical studies associate higher levels of earnings volatility with fair value accounting (Barth et al. 1995; Hodder et al. 2006). In contrast, the author found evidence that banks applying the fair value measurement to reduce accounting mismatches exhibit lower earnings volatility than other banks. He concludes that banks can use the flexibility in accounting to reduce artificial earnings volatility.

Same results have been achieved by Dechow and Shakespeare (2009) who documented that firms time securitization transactions to suit their financial reporting purpose. Moreover, Dechow et al. (2010) found evidence that securitization gains are significantly negatively related to pre-securitization gains. Together, those studies indicate that securitizations are employed as an earnings management tool, either through 'real' earnings management (timing of the gains recognition) or through the discretion over accounting assumptions. So that, this results indicate that managers use the flexibility available in fair value accounting rules to smooth earnings.

Answering to Dechow et al. (2010) and in defense of fair value, Barth and Taylor (2010) clarify the role of fair value in accounting for asset securitizations, discussing alternative explanations for the evidence presented in DMS, and offering suggestions for future research, paying caution against inferring the desirability of any particular accounting method from earnings management research. 'Real' earnings management, following Barth and Taylor (2010), refers to situations where firms enter into transactions that alter current period earnings, but do *not* manipulate the accounting estimates. The discretion over the timing of securitizations puts securitization transactions in this category. Real earnings management smoothes out accounting earnings, but it does *not* really undermine the integrity of securitization accounting (Dechow and Shakespeare, 2009).

Building on Henry's (2009) study of early adopting banks, the paper of Guthrie et al. (2011) examines to what extent firms' election of instruments benefited their current or future earnings. Under the fair value option, SFAS 159, firms have full discretion over electing to report specified financial instruments at fair value on a contract-by-contract basis. Sample adopted comprises the constituents of the S&P 1500 Index for the first quarters of fiscal years 2007 and 2008. Expanding the sample across industries and over time allows the authors to obtain a more complete picture of the adoption of the fair value option. The authors do not find evidence of systematic opportunistic election of the fair value option. In only a handful of cases, concentrated among early adopters with an earnings shortfall, did firms experience a significant improvement in current or future earnings that casts doubt on whether their adoption was keeping with the intent and spirit of the standard.

Focusing on topic related to earning management, Goncharov et al. (2011) examine the impact of positive fair value adjustments on dividend policy. Authors state that, if fair value adjustments are transitory in nature and managers are able to assess their implications for future earnings, fair value adjustments in net income is expected to have no distribution consequences. However, positive fair value adjustments may lead to higher dividends when management incorrectly assesses their persistence, this having a potential for pro-cyclical impact

because higher dividends increase leverage, and thus risk. Finally, they found no empirical support for the concern that dividends increase in response to positive fair value adjustments.

Considering advantage and disadvantages of FVO, Magnan (2009) discusses how FVA affects the nature of financial reporting, especially for financial institutions that were deeply affected by the 2007-9 financial crisis. The evidence of the investigation does suggest that FVA, in combination with its use by regulators, may have severely undermined the financial condition of some institutions. In particular, the effect was amplified for institutions holding assets in markets that saw their liquidity dry up during the crisis. In other words, FVA may have amplified the crisis.

On the topic, Kothari et al. (2010) review the positive theory of GAAP that predicts that GAAP's principal focus is on control (performance measurement and stewardship) and that verifiability and conservatism are critical features of a GAAP shaped by market forces. The authors recognize the advantage of using fair values in circumstances where these are based on observable prices in liquid secondary markets, but caution against expanding fair values to financial reporting more generally. They conclude that rather than converging U.S. GAAP with IFRS, competition between the FASB and the IASB would allow GAAP to better respond to market forces

Considering the impact of fair-value accounting on financial statement analysis, Rodríguez-Pérez et al. (2011) tried to shed some light on this issue by restating the financial investments and tangible fixed assets of a sample of 85 Spanish insurance companies, applying fair value instead of historical-cost-based valuations and by simulating analyst perception of these companies' efficiency and profitability for both sets of data using data envelopment analysis (DEA), a method used to empirically measure productive efficiency of decision making units. They found that the numbers on the face of the financial statements change considerably and observe that the magnitude of these changes varies between companies and classes of assets. However, only in a few cases does a change in the valuation basis lead to a relevant change in DEA scores; within the sample, the overall assessment of companies with regard to efficiency and profitability remains largely the same under both valuation bases. Findings of Rodríguez-Pérez et al. (2009) seem to indicate that a change from historical-cost to fair-value accounting could alter analyst perceptions of a limited number of companies but likely will not have a major impact on the appraisal of the majority of them.

Given the previous debate, the aim of the paper is to verify if the principles adopted by the Boards referring to Fair Value Hierarchy are effective in practice and if this principles permit to obtain a clear disclosure of the value and the risks joined to the financial instruments owned by the bank sector. Moreover, the analysis of data is completed with the construction of some regressive equations with the aim to test the relations between some variables described in financial statement and the volume, the evaluation models and the information referring to the different level of fair value existing for financial instruments – as it will better described in the following paragraph. From the qualitative and the quantitative analysis is also possible to obtain a first indication referring to the utility of this data for the investors.

3. Reasons of the empirical analysis

The validity of the theory on different relations existing between accounting information released in the Annual Reports of a sample of listed banks, market capitalization at year end and the three levels of fair value adopted to assess the value of assets and liabilities (Fair Value Hierarchy) has been tested. For the purpose of the study we used statistical tools as multivariate regression analysis. In particular, the aims of the analyses are:

i. To verify if a significant relation can be found between specific accounting information reported on the Annual Report of a company in the sample and a set of specific variables related to each company. In particular, we focused on the explanatory power of three levels of fair value adopted to assess assets and liabilities - related to fair value hierarchy to explain dynamics market related (i.e., market capitalization) and accounting variables (i.e., net income) over time;

- ii. to verify if it exists a relation between the annual percentage performance of portfolios of assets (or the net performance considering also liabilities) evaluated at fair value of level one and the annual percentage performance of portfolios of assets (or the net performance considering also liabilities) evaluated at fair value of level two and three;
- iii. to verify whether the relationships detected in (i) and (ii) are statistically significant and have a good explanatory power;
- iv. To interpret the quantitative results in order to draw conclusions about the impact of each fair value measure on the selected dependent variables in a corporate governance perspective analysis.

Furthermore, the study is thorough for detecting eventual homogeneities among *sub*-groups of companies characterized by common characteristics in terms of geographic area and quantitative relevance of the third level-fair value measure of assets.

4. Methodology and data design

1. Data source of the analysis

Since the raw data used to calculate each explanatory variables (EV) – such as FV x A and FV x L, for each level of fair value one, two and three - and part of the dependent variable (DV) as Net Income (NI) are calculated on the basis of the information reported on the companies' Annual Reports, for each year object of study the model estimates are referred to the closing date of Annual Report at December 31^{st} .

The models shown on Table 2.2 are tested for the years 2009, 2010 and 2011 considering the introduction of Fair Value Hierarchy as stated in accounting principles. As some variables are expressed in terms of annual variation, data have been collected also for 2008. The data have been retrieved on the economic-financial database Bloomberg.

The raw data used to determine the value of the variables described in Table 2.1 are referred to a sample of companies (the "Sample") selected through the following criteria:

- i) Companies that are not listed on any equity stock market have been excluded from the sample. This criteria is necessary as we need to calculate the annual stock market price variation of firms;
- ii) for each listed firm only the primary equity security is taken into account in order to avoid to consider more than once the same company;
- iii) the selection is limited to the companies which operated in banking sector (Bloomberg ICB sector "Banks") in the three-year period considered;
- iv) we include in the selection only the companies for which, on their last official document available on Bloomberg, at the reference date of the research, it was specified the information regarding assets' fair value measures of each level (one, two and three).
- v) The sample was increased to include some specific Italian banks that were not found by Bloomberg through the research criteria (function <EQS>) specified in the previous points from i) to iv).

Finally, the availability of data, *de facto*, has operated as additional criteria which excluded from the Sample the companies whose raw data, at the research date, were not available, entirely or in part, on the database used to collect information.

The Sample identified by the intersection of these criteria is reported in Table A.5; however, the amplitude and the composition of the sample used for the analyses vary for each regression depending on the availability of the specific data which are necessary to calculate the DV and the EV of the model.

Furthermore, different subsets of the Sample have been identified to test the validity of the model also for sub-groups of banks characterized by homogeneities in terms of:

a. Geographic area of the country of Head Quarter.

The Sample is subdivided into three groups: World (all the Sample), Italy and United States. This subdivision is necessary because, even if IASB and FASB tried – in the recent periods - to converge referring to the fair value accounting principles, some differences. Moreover, some countries apply local GAAP instead of IFRS and SFAS. The subdivision permits us to verify if there are differences in fair value hierarchy disclosure joined to the nationality of the banks;

b. Quantitative relevance of the third level-fair value measure of assets (FV 3 A).

For each model, year (t) and observed company (i) we defined the following indicator ("FV 3 %R" in equation 3.1), which expresses the quantitative relevance of the third level-fair value measure of assets on the total sum of fair value measures of assets referred to company (i) at time (t):

$$FV \ 3 \ \% R_t^i = \frac{FV \ 3 \ A_t^i}{FV \ 1 \ A_t^i + FV \ 2 \ A_t^i + FV \ 3 \ A_t^i}$$
(3.1)

Afterwards, for each *sub*-sample examined, we calculated the 33,33% (P₃₃) and the 66,67% (P₆₆) percentile of FV 3%R empirical distribution, defining three possible ranges of the indicator:

- first percentile range: $FV \ 3 \% R_t^i \le P_{33}$;

second percentile range: $P_{33} < FV \ 3 \% R_t^i \le P_{66}$;

- third percentile range: $FV \ 3 \% R_t^i > P_{66}$.

Finally, the sub-sample's raw data and variables were divided into three groups:

- i) PERC. FV 3 n°1: the companies whose FV 3%R indicator value fell in the first percentile range;
- ii) PERC. FV 3 n°2: the companies whose FV 3%R indicator value fell in the second percentile range;
- iii) PERC. FV 3 n°3: the companies whose FV 3%R indicator value fell in the third percentile range.

Before testing hypotheses and running regressions, a qualitative analysis of data available has been performed in order to verify if data provided in financial statement (i) are compliant to current accounting principles (ii) are clear, understandable and useful for stakeholders/investors purpose. This step of the analysis provided us useful information for the purpose of this study but, due to the big volume of data, results are not shown. Nevertheless, this step helped us to select only the set of companies formally compliance with accounting principles.

2. Set of variables

The set of dependent variables selected – *Net Income* and *Market Capitalization* – are highly significant for investors and external stakeholders. These two variables, even if they can suffer of potential effect of other factors (i.e., accounting policy for Net Income; short term market conditions for Market Capitalization) they can be considered good proxies to set up the economics of a company. Regarding the independent variables, different levels of fair value and annual changes in value are considered, these representing key elements for the purpose of our study.

From a technical point of view, the variety of possible explanatory variables ("EV") that can be defined with reference to each level of fair value and, similarly, the multiplicity of dependent variables ("DV") that could be linked to those fair value measures, has determined the need to test a more than one statistical model. In particular, Table 2.1 shows the set of DV and EV object of our analyses.

Table 2.1 - Dependent variables and explanatory variables object of analysis

DV - Dependen	t variables	
Symbol	Description	Calculation formula
MC	Market capitalisation	Raw data
NI	Net income	Raw data
EV - Explanato	ry variables	
Symbol	Description	Calculation formula
FV x A	Fair value measure of level (x) of assets	Raw data
FV x L	Fair value measure of level (x) of liabilities	Raw data
Δ FV x A	Absolute variation between year [t-1] and year [t] of FV x A	$(FV \times A [t]) - (FV \times A [t-1])$
Δ FV x L	Absolute variation between year [t-1] and year [t] of FV x P	(FV x L [t]) - (FV x L [t-1])
FV x net	Net fair value measure of level (x)	(FV x A) - (FV x L)
Δ FV x net	Absolute variation between year [t-1] and year [t] of FV x net	$(\Delta \text{ FV x A})$ - $(\Delta \text{ FV x L})$
Δ% FV x A	Percent change between year [t-1] and year [t] of FV x A	Δ FV x A / (FV x A [t-1])
Δ% FV x net	Percent change between year [t-1] and year [t] of FV x net	Δ FV x net / (FV x net [t-1])

3. The relationship investigated: models

The following Table 2.2 shows the structure – in terms of DV and EV - of the statistical models examined in this paper.

Table 2.2 – The structure of the examined statistical models in terms of DV and EV.

Model n°	Dependent variable		Independent variables	
#	DV	EV.1	EV.2	EV.3
1	MC	FV 1 A	FV 2 A	FV 3 A
2	NI	Δ FV 1 A	Δ FV 2 A	Δ FV 3 A
3	Δ% FV 1 A	-	$\Delta\%$ FV 2 A	$\Delta\%$ FV 3 A
4	Δ% FV 1 net	-	$\Delta\%$ FV 2 net	$\Delta\%$ FV 3 net

In order to refine the analysis, tests for four different models have been provided. For each model tested, the coefficients have been estimated adopting the Ordinary Least Squares ("OLS") method through a multivariate linear regression.

In formula (the "Model" of equation 2.1):

$$DV_t^i = \beta_0 + \beta_1 EV. \mathbf{1}_t^i + \beta_2 EV. \mathbf{2}_t^i + \beta_3 EV. \mathbf{3}_t^i + \varepsilon_t^i$$
(2.1)

The hypotheses at the base of the Model are the standard assumptions of the OLS. As specified in paragraph 4, some of them have been tested in order to verify if this method is applicable to the data set for the purpose of the analyses.

Since the models are cross sectional, the value of DV_t^i is estimated for a reference date (t) as a function of the value that the independent variables assume at the same time (t). Parameters have been estimated for a fixed

value of (t) on the basis of the observations of the variables related to the (n) elements of the selected sample of companies described in paragraph 1.

Main objective of the analyses is to set if a stable and direct relation can be assessed between different level and annual changes of fair value, as stated in accounting principles, and two main variables related to companies, net income as reported in annual statement and market capitalization at year end.

4. The tests

The models of Table 2.2 are singularly tested varying the reference time (t) and the *sub*-sample of companies considered developing the OLS regressions as described above.

For a given model of Table 2.2, considering the *sub*-divisions specified in point a. (three geographic areas) and in point b. (three ranges of FV 3 A) of paragraph 3 and that we repeated these analyses for each of the selected years (2009, 2010 and 2011) we have developed a total of eighteen regressions per model. Consequently, for examining all the models of Table 2.2 we have run 72 OLS regressions.

For each regression, in addition to the analysis of coefficients values, we checked the statistical significance of the model through the following tests:

• F statistic

To verify the statistical significance of the whole model;

Student's t

To verify the statistical significance of the single explanatory variables;

Adjusted R-squared

To measure the ability of the model to explain the variance of sample observations.

For each model of Table 2.2, the study of each combination [year (t); *sub*-sample of companies] is structured in two phases:

- i) The parameters of the Model are estimated including all the EV specified in Table 2.2. Therefore, we can verify the singular statistical significance of each independent variable through its Student's t P-Value;
- ii) On the base of the results of phase i), a second-step OLS regression is repeated for the same combination [year (t); *sub*-sample of observations] excluding the independent variables whose Student's t in the first regression resulted statistically not significant at a level of confidence of 95% (P value ≤ 5%). Considering all the scenarios, the final number of regressions run in the first and in the second phase is equal to 144 − that is;

$$144 = \{ \left[\left(3_{geogr. \; areas} + \, 3_{FV \; 3 \; percentiles} \right) \cdot 3_{years} \right] \cdot 4_{models} \} \, \cdot \, 2_{analysis \; phases}.$$

Since the null hypothesis (coefficient of the single explanatory variable equal to zero) cannot be rejected at a level of confidence of 95% for Student's t P-values higher than 0,05, for the purposes of this paper we consider only the results of the second-step analyses described in ii).

5. Empirical evidence on testing hypothesis: results and discussions

All the models shown in Table 2.2 have been tested through both the first-step and the second-step groups of regressions described in the previous paragraph.

Considered that each model has been tested through eighteen different regressions and considered the plurality of statistical indicators used to verify the validity of each of them, the general judgment of statistical

significance has been characterized by inevitable elements of subjectivity. Furthermore, for some models (3 and 4), the results of the analyses show a low statistical significance but anyway they are relevant to draw conclusions in the perspective of a corporate governance study.

In general, with reference to Model 1 and 2, the outputs of the regressions (shown in Appendix, Table A.1 and Table A.2) evidence that, on average, the F statistic assumes very high values for most of the dates (t) and the *sub*-samples analyzed, showing that, in most of the cases, coefficients are jointly significant independently from the reference year and from the specific companies considered in the regression.

Similarly, also the adjusted R-squared, on average, is very high for all the 36 regressions relating to Model 1 and 2. In particular, it's interesting to notice that its average value is 76,23%, while the median one is 86,43%.

As it is reasonable to expect, the proportion of variability in the sample observations that is accounted for by the explanatory variables (measured by the adjusted R-squared) increases when we consider a more restricted geographic area. For each year the statistic is higher for USA and Italy, while it's significantly lower – despite still high in absolute – for the whole Sample. This can be reasonably be attributed to the fact that for groups of companies characterized by the same country of domicile the values of coefficients can be better estimated in order to reflect the nation-specific social, economic, financial, and cultural factors, thus allowing a more accurate estimate of bank's book value.

On the contrary, it is interesting to notice that, for each of the three years, on average, the maximum values of Adjusted R-squared are observable in the first percentile range, following in decreasing order the second and the third percentile range.

A common characteristic observed among the results of the analysis of significant models is the instability of the independent variables structure.

The following analyses examine each model in order to detect eventual regularities in the EV structure and to draw conclusions about the characteristics concerning the corporate governance of companies across the different *sub*-samples examined.

Hypothesis 1: There is any statistical and significant correlation between market capitalization at year end and financial asset evaluated at different level of fair value?

$$MC_t^i = \beta_0 + \beta_1 (FV \ 1 \ A)_t^i + \beta_2 (FV \ 2 \ A)_t^i + \beta_3 (FV \ 3 \ A)_t^i + \varepsilon_t^i$$

The results of the regression for Model 1 are presented in Appendix in Table A.1.

Similarly to Model 1, on average, USA presents the highest level of significance in terms of both F statistic and single Student's t of coefficients. Moreover, it is characterized by the highest values of adjusted R-squared.

Both for Italy and USA the third level measure of fair value of assets (FV 3 A) is significant to estimate the market capitalization of banks for all the years examined.

With reference to the *sub*-samples distinguished according to the percentile range of FV 3% R, we notice that, on average, FV 3 A is the explanatory variable which results statistically significant in most of the cases, followed by FV 2 A and FV 1 A.

Model 1 shows that market capitalization of banks in the sample seems to have an high level of correlation with asset evaluated at fair value level three, although these results cannot be generalized looking at different level of significance considering three subsamples (World, Usa, Italy). It's interesting to note that in US market different level of fair value in the valuation of asset for US market seems to be always significant. Otherwise, in Italian market significance of fair value asset at different level is higher over time (low in 2009, high in 2011).

This evidence can allow us to say that Italian market is going to improve the adoption of fair value option to better assess value of asset portfolio.

Focusing on Italian market, results show that FV asset level three is always significant, this indicating that banks can use fair value option at level three to mitigate negative effect of undervaluation due to particular market conditions, as in the period 2009-201 after financial crisis.

Hypothesis 2: There is any statistical and significant correlation between net income at year end and annual change in asset evaluated at different level of fair value?

$$NI_t^i = \beta_0 + \beta_1 (\Delta FV 1 A)_t^i + \beta_2 (\Delta FV 2 A)_t^i + \beta_3 (\Delta FV 3 A)_t^i + \varepsilon_t^i$$

The results of the regression for Model 2 are presented in Appendix in Table A.2.

We estimated NI as a function of the annual variation of the different level of fair value measures of assets. The results show that USA is characterized again by the highest values of F statistic and adjusted R-squared, but the independent variables structure is unstable across the years.

In the percentile ranges differentiation of *sub*-samples, the EV structure is highly unstable and in 33,3% of cases the coefficients result to be jointly not significant, hence we reject the hypothesis of significant differences among the groups of banks characterized by different levels of FV 3 impact on the total of fair value measures of assets.

In general, the regressions output of Model 2 evidences an interesting homogeneity across the different *sub*-samples analyzed: in most of the cases the coefficient of Δ FV 3 A is negative, while the coefficient of Δ FV 1 A is positive. The second level-measure, Δ FV 2 A, is placed at an intermediate point.

This result can allows us to affirm that fair value option, in particular considering FV level 3, can be considered as, among others influencing net income value, one of the tool to mitigate effect of countercyclical trend in bad years such as ones observed. Nevertheless, this proposition has to be tested more in deep, considering other factors influencing net income dynamics.

For this purpose, with reference to the same companies considered to run the regressions of the World *sub*-sample in Model 2, it is interesting to analyze the empirical percentages of times in which, for each combination of year (t) and level (x) of fair value, the value of the variable Δ FV x A has the same sign of the net income (shown in Table 5.2).

Table 5.2 – Empirical percentages of times in which, for each combination of year (t) and level (x) of fair value, the value of the variable Δ FV x A has the same sign of NI.

Year	Δ FV 1 A	Δ FV 2 A	Δ FV 3 A
2009	60,42%	54,17%	45,83%
2010	57,14%	61,90%	33,33%
2011	47,89%	61,97%	46,48%

It is evident to notice that Δ FV 1 A and Δ FV 2 A are characterized by the highest percentages, while those referred to Δ FV 3 A are significantly lower. Furthermore, the lowest percentages of Δ FV 3 A (in absolute) and the widest differentials with between the third level and the first two ones are observable for 2009 and 2010, namely the years that immediately followed the financial crisis which begun in 2008. This can be related to the hypothesis that FV3 can be used as countercyclical tool for earning management.

Similar results are obtained repeating the same analyses with reference to the net measure of fair value (Table 5.3).

Table 5.3 – Empirical percentages of times in which, for each combination of year (t) and level (x) of fair value, the value of the variable Δ FV x net has the same sign of NI.

Year	ΔFV 1 net	ΔFV 2 net	ΔFV 3 net
2009	62,96%	51,85%	37,04%
2010	55,26%	55,26%	39,47%
2011	48,89%	57,78%	48,89%

Before any definitive conclusion, the study should be further investigated in order to remove the assumptions described above and to consider also the effect of the other variables that influence NI. However, the results of our analyses constitute evidence that, with reference to our Sample of banks, in the years from 2009 to 2011 the variables calculated as a function of the third level-measure of fair value behaved in a way that was significantly different, in statistical terms, in comparison to the ones calculated as a function of the first and the second-level measures of fair value.

We proceed in the analysis of Model 3 (4), in which the annual percentage change of FV x A (FV x net) is expressed as a function of the annual percent change of the same variable referred to the second and the third level measure of fair value. Hence, in comparison to the models previously analyzed, the number of repressors is reduced to two.

The aim of these models is to verify if the second and the third level-measures of fair value vary in accordance to the first one, or if they vary in a significantly different way, or if there is not any significant linear relationship. In this sense, we conjecture that, as FV 1 A (FV 1 net) are a *proxy* of the market indexes that must be taken as a reference for evaluating assets and liabilities of the second and third level of fair value independently from the specific models used by banks for their assessment -, Δ % FV 2 A (Δ % FV 2 net) and Δ % FV 3 A (Δ % FV 3 net) should vary in accordance to Δ % FV 1 A (Δ % FV 1 net).

In addition to the assumptions specified above, models 3 and 4 require a further hypothesis: for each level of fair value, the portfolios of assets and liabilities assessed at fair value are characterized by the same internal distribution of financial instruments in terms of typology and economic value. Since we focus on the annual percentage changes, no hypothesis is made on the absolute total value of each portfolio, but only on their internal qualitative composition and the percentage weight of the categories of elements that constitute them. As for the previous assumptions, also this hypothesis could be removed in further studies examining the internal composition of each level of fair value-measure.

Hypothesis 3: There is any statistical and significant correlation between annual change in fair value asset of first level and annual change in fair value asset at level two and level three?

$$(\Delta\% FV \ 1 \ A)_t^i = \beta_0 + \beta_2 (\Delta\% FV \ 2 \ A)_t^i + \beta_3 (\Delta\% FV \ 3 \ A)_t^i + \varepsilon_t^i$$

The results of the regression for Model 3 are presented in Appendix in Table A.3.

It is evident to notice that the model is not significant in most of the cases, as in 10 regressions out of 18 we cannot refuse the hypothesis that coefficients are jointly equal to zero. With reference to the cases of joint significance of coefficients, we don't detect any stability in the EV structure across the years and the *sub*-groups of companies examined.

Consequently, we can infer that no relevant linear relationships exist between the percentage performance of FV 1 A and the percentage performance of FV 2 A and FV 3 A.

Hypothesis 4: There is any statistical and significant correlation between annual change in net fair value of level 1 and annual change in net fair value at level two and level three?

$$(\Delta\%~\textit{FV}~1~\textit{net})_t^i = \beta_0 + \beta_2 (\Delta\%~\textit{FV}~2~\textit{net})_t^i + \beta_3 (\Delta\%~\textit{FV}~3~\textit{net})_t^i + \varepsilon_t^i$$

The results of the regression for Model 4 are presented in Appendix in Table A.4. Unlike Model 3, for each level of fair value we express DV and the EV in terms of net fair values.

The analysis of Model 4 has confirmed also for the net measures of fair value the results obtained with reference to Model 3 about the non-correlation between the annual percentage performance of the financial instruments assessed at fair value of level one and the percentage performance of those assessed at fair value of second and third level.

6. Conclusion and further research

The paper investigate if any relation between fair value hierarchy and variables related to market capitalization and net income can be assessed, assuming that data used in the analysis are formally complaint to international accounting principles. Moreover, a second level of analysis tries to evaluate if any relation can be found between changes in value of FV1 asset and FV1 net (defined as accounting value of asset minus accounting value of liabilities) can be found.

In the first part of the analysis, we found that market capitalization and net income are quite correlated to value of asset evaluated at different level of fair value. This is evidence is particularly strong for subsample US market. This allows us to say that, even if Us Gaap and IFRS 13 can be considered quite close in evaluation of financial asset through fair value option, the evidence have to be investigated more in deep.

Looking at relation between net income and changes in value of asset evaluated at different level of fair value, results show that level three of fair value can be considered as a countercyclical tool available to be used in contrast in bad period, such as 2009-2011 characterized by financial crisis.

Considering models adopted, results from our analysis can be summarized as follows:

- Fair value level 3, that is the more subjective criteria in financial instruments measurement, shows poor relevance in US banks and much less relevance in Italy;
- Disclosure on fair value hierarchy is not widely adopted: considering a potential sample of more than 2500 listed banks, only 281 currently disclose on three different level of fair value, as requested by accounting principles (IFRS 7 and SFAS 157);
- Even if financial instruments assessed adopting fair value level 3 are quite dissimilar in value considering level 1 and 2 of fair value, a deeper disclosure seems to be required for this class;
- Since data on fair value level 3 show results not ever consistent in time and space, it seems to be necessary to better investigate in order to assess if "anomalies" can be referred to specific class of financial asset, market trend, models and assumptions adopted for evaluation;
- Even if the fair value level 3 is more subjectivity, the degree of subjectivity in evaluation of financial instruments of level 2 more significant in value considering the whole portfolio has to be taken into account.

Results of regression analysis show that variables investigated – market capitalization, net income and three level of fair value – are quite associated, under specific assumptions, but they do not offer unique and clear information to investors in terms of usefulness for their capital allocation strategy.

In synthesis, even if fair value hierarchy principles allow a better understanding about trend in value and composition of financial instruments portfolio of banks, they suffer of two main limitations: (i) subjectivity problem in value estimation; (ii) short term volatility in results due to changes in macroeconomic variables.

7. Limitations of the analysis

The analyses are based on the assumption that, for each year and company object of study, the portion of assets and liabilities assessed at fair value that have been reclassified at a different fair value level from one year to the next is equal to zero. This coincides with the assumption that the annual change of level (x)-fair value measure is entirely attributable to the variation of value of the assets and the liabilities assessed at that level (x) and not to a change in the valuation criteria (to a different fair value level) of financial instruments assessed at fair value. Furthermore, we assume that, for each level of fair value, companies did not increase nor decrease from one year to the next the amount of resources invested (for assets) and borrowed (for liabilities) that are assessed at fair value. Thus we assume that companies may changed portfolios composition, but did not disinvest nor invest new resources in fair value-assessed financial instruments from one year to the next.

Both hypotheses could be removed in further works investigating, for each level of fair value, the portfolios composition of assets and liabilities assessed at fair value.

In synthesis, hypotheses at the base of the models investigated have to be assessed to better fit the complexity of the economics involved. In fact, even if results show a good degree of correlation between market capitalization and net income, the study has to be improved to take into account some main aspects

- (i) composition of each portfolio of asset and liabilities,
- (ii) change in composition an portfolio
- (iii) Specific weight of each class of asset and liabilities considering the whole value of asset portfolio and financial structure of bank' sample.

These limitations are relevant in particular for models 3 and 4 to fix any conclusions that can allow us to affirm that a good degree of correlation can be found and that banks use fair value option to mitigate effect of bad years in terms of portfolio performance and, hence, in terms of earnings management.

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Appendix

Table A.1 - Parameters estimation and tests of significance for Model 1.

Model 1 - regr. (ii)		2009			2010			2011	
Sub-sample	WORLD	ITALY	USA	WORLD	ITALY	USA	WORLD	ITALY	USA
Number of observations	186	13	146	211	14	166	240	15	191
F statistic	256,874	1.483,471	689,601	356,073	982,108	2.381,461	912,064	234,009	2.104,116
P-Value F statistic	0,000	0,000	0,000	0,000	0,000	0,000	0,000	0,000	0,000
Adjusted R- squared	72,94%	90,86%	93,44%	77,18%	91,01%	97,74%	88,40%	89,71%	95,68%
Significant EV									
Intercept	-	-	973.953,6	41.727.992,5	-	1.088.451,7	23.781.943,4	-	1.020.241,0
P-Value Intercept	-	-	0,045	0,026	-	0,000	0,022	-	0,000
FV 1 A	0,260	-	-0,272	4,962	0,198	-0,463	4,123	0,245	-0,659
P-Value FV 1 A	0,000	-	0,000	0,000	0,000	0,000	0,000	0,000	0,000
FV 2 A	-	-	-0,047	-0,541	-	-0,048	-0,454	-0,132	-
P-Value FV 2 A	-	-	0,000	0,000	-	0,000	0,000	0,001	-
FV 3 A	0,573	0,542	2,485	-	1,269	3,807	-	1,442	2,900
P-Value FV 3 A	0,000	0,000	0,000	-	0,000	0,000	-	0,001	0,000
Sub-sample	PERC. FV 3 n°1	PERC. FV 3 n°2	PERC. FV 3 n°3	PERC. FV 3 n°1	PERC. FV 3 n°2	PERC. FV 3 n°3	PERC. FV 3 n°1	PERC. FV 3 n°2	PERC. FV 3 n°3
Number of observations	61	64	61	68	72	71	81	80	79
F statistic	1.032,997	197,169	284,926	5.963,351	175,929	318,604	376,345	1.238,714	273,744
P-V alue F statistic	0,000	0,000	0,000	0,000	0,000	0,000	0,000	0,000	0,000
Adjusted R-	00.400/						02 000/	05 (00)	86,21%
squared	98,10%	84,58%	80,94%	98,89%	86,65%	80,56%	92,09%	95,63%	· · · · · · · · · · · · · · · · · · ·
squared Significant EV	98,10%	84,58%	80,94%	98,89%	86,65%	80,56%	92,09%	95,63%	
1	963.239,1	84,58%	80,94%	-891.195,1	86,65%	80,56%	92,09%	95,63%	-
Significant EV				,			92,09%	95,63%	-
Significant EV Intercept	963.239,1	-		-891.195,1	-		92,09% - - -2,739	95,63%	-
Significant EV Intercept P-Value Intercept	963.239,1 0,019	-	-	-891.195,1 0,000	-	-	-	95,63%	- - -
Significant EV Intercept P-Value Intercept FV 1 A P-Value FV 1 A FV 2 A	963.239,1 0,019 1,117	- 0,586	-	-891.195,1 0,000	1,812	4,460	-2,739	95,63%	- - - -0,089
Significant EV Intercept P-Value Intercept FV 1 A P-Value FV 1 A	963.239,1 0,019 1,117 0,000	- - 0,586 0,000	-	-891.195,1 0,000	- - 1,812 0,000	- - 4,460 0,000	-2,739 0,000	- - -	- - -
Significant EV Intercept P-Value Intercept FV 1 A P-Value FV 1 A FV 2 A P-Value FV 2	963.239,1 0,019 1,117 0,000 -0,154	- 0,586 0,000	-	-891.195,1 0,000 - - 1,357	1,812 0,000 0,509	4,460	-2,739 0,000 0,832	- - - -3,435	- - - -0,089

Table A.2 - Parameters estimation and tests of significance for Model 2.

Model 2 - regr. (ii)		2009			2010			2011	
Sub-sample	WORLD	ITALY	USA	WORLD	ITALY	USA	WORLD	ITALY	USA
Number of observations	48	9	25	63	9	27	71	14	28
F statistic	41,859	112,136	2.150,321	31,261	60,318	3.106,881	35,720	103,521	2.811,524
P-Value F statistic	0,000	0,000	0,000	0,000	0,000	0,000	0,000	0,000	0,000
Adjusted R- squared	46,51%	82,26%	95,10%	49,40%	75,79%	95,58%	58,57%	81,15%	99,52%
Significant EV	17120211			4 454 450 0					166 600 0
Intercept P-Value Intercept	4.712.934,4	-	-	4.674.159,2	-	-	-	-	466.620,3
Δ FV 1 A	0,024	0,470	0,255	0,005	0,454	1,907	1,011	-	0,028
P-Value Δ FV 1	-	0,000	0,002	-	0,000	0,000	0,000	-	-
Δ FV 2 A	-0,054	-1,192	-0,191	-0,197	-	-	0,156	-	0,197
P-Value ∆ FV 2 A	0,000	0,000	0,000	0,000	-	-	0,004	-	0,000
Δ FV 3 A	-	-	-	-2,797	-	-3,698	-2,139	12,620	-0,508
P-V alue Δ FV 3 A	-	-	-	0,000	-	0,000	0,016	0,000	0,001
Sub-sample	PERC.	PERC. FV	DEDC EV	DEDC	DED C EV	DEDC EV	PERC.	PERC. FV	DEDC EV
Sub-sample	FV 3 n°1	3 n°2	PERC. FV 3 n°3	PERC. FV 3 n°1	PERC. FV 3 n°2	PERC. FV 3 n°3	FV 3 n°1	3 n°2	PERC. FV 3 n°3
Number of		3 n°2 NOT	3 n°3 NOT				FV 3 n°1		
	FV 3 n°1	3 n°2	3 n°3	FV 3 n°1	3 n°2	3 n°3	FV 3 n°1	3 n°2	3 n°3
Number of observations	FV 3 n°1	NOT SIGNIF.	3 n°3 NOT SIGNIF.	FV 3 n°1	3 n°2	3 n°3	FV 3 n°1 NOT SIGNIF.	3 n°2	3 n°3
Number of observations F statistic P-Value F	FV 3 n°1 10 89.258,102	NOT SIGNIF.	3 n°3 NOT SIGNIF.	10 5.641,276	3 n°2 22 29,142	3 n°3 31 32,895	FV 3 n°1 NOT SIGNIF.	3 n°2 22 148,881	3 n°3 33 42,296
Number of observations F statistic P-Value F statistic Adjusted R-	FV 3 n°1 10 89.258,102 0,000	NOT SIGNIF.	3 n°3 NOT SIGNIF.	FV 3 n°1 10 5.641,276 0,000	22 29,142 0,000	3 n°3 31 32,895 0,000	FV 3 n°1 NOT SIGNIF.	3 n°2 22 148,881 0,000	3 n°3 33 42,296 0,000
Number of observations F statistic P-Value F statistic Adjusted R-squared Significant EV Intercept	FV 3 n°1 10 89.258,102 0,000	NOT SIGNIF.	3 n°3 NOT SIGNIF.	FV 3 n°1 10 5.641,276 0,000	3 n°2 22 29,142 0,000 72,83% 4.394.140,6	3 n°3 31 32,895 0,000 51,53% 4.860.917,6	FV 3 n°1 NOT SIGNIF.	3 n°2 22 148,881 0,000	3 n°3 33 42,296 0,000
Number of observations F statistic P-Value F statistic Adjusted R-squared Significant EV Intercept P-Value Intercept	FV 3 n°1 10 89.258,102 0,000 87,49%	3 n°2 NOT SIGNIF	3 n°3 NOT SIGNIF. -	FV 3 n°1 10 5.641,276 0,000 88,73%	3 n°2 22 29,142 0,000 72,83% 4.394.140,6 0,033	3 n°3 31 32,895 0,000 51,53% 4.860.917,6 0,073	FV 3 n°1 NOT SIGNIF.	3 n°2 22 148,881 0,000 90,23%	3 n°3 33 42,296 0,000 69,09%
Number of observations F statistic P-Value F statistic Adjusted R-squared Significant EV Intercept P-Value Intercept Δ FV 1 A	FV 3 n°1 10 89.258,102 0,000 87,49%	NOT SIGNIF.	3 n°3 NOT SIGNIF. -	FV 3 n°1 10 5.641,276 0,000	3 n°2 22 29,142 0,000 72,83% 4.394.140,6	3 n°3 31 32,895 0,000 51,53% 4.860.917,6	FV 3 n°1 NOT SIGNIF.	3 n°2 22 148,881 0,000 90,23%	3 n°3 33 42,296 0,000
Number of observations F statistic P-Value F statistic Adjusted R-squared Significant EV Intercept P-Value Intercept	FV 3 n°1 10 89.258,102 0,000 87,49%	3 n°2 NOT SIGNIF	3 n°3 NOT SIGNIF. -	FV 3 n°1 10 5.641,276 0,000 88,73%	3 n°2 22 29,142 0,000 72,83% 4.394.140,6 0,033	3 n°3 31 32,895 0,000 51,53% 4.860.917,6 0,073	FV 3 n°1 NOT SIGNIF.	3 n°2 22 148,881 0,000 90,23%	3 n°3 33 42,296 0,000 69,09%
Number of observations F statistic P-Value F statistic Adjusted R-squared Significant EV Intercept P-Value Intercept Δ FV 1 A P-Value Δ FV 1 A Δ FV 2 A	FV 3 n°1 10 89.258,102 0,000 87,49% -0,703	3 n°2 NOT SIGNIF	3 n°3 NOT SIGNIF.	FV 3 n°1 10 5.641,276 0,000 88,73% 0,804	3 n°2 22 29,142 0,000 72,83% 4.394.140,6 0,033	3 n°3 31 32,895 0,000 51,53% 4.860.917,6 0,073	FV 3 nº1 NOT SIGNIF.	3 n°2 22 148,881 0,000 90,23% - 0,836	3 n°3 33 42,296 0,000 69,09% 0,257
Number of observations F statistic P-Value F statistic Adjusted R-squared Significant EV Intercept P-Value Intercept	FV 3 n°1 10 89.258,102 0,000 87,49% -0,703 0,000	3 n°2 NOT SIGNIF	3 n°3 NOT SIGNIF.	FV 3 n°1 10 5.641,276 0,000 88,73% - 0,804 0,000	3 n°2 22 29,142 0,000 72,83% 4.394.140,6 0,033 -	3 n°3 31 32,895 0,000 51,53% 4.860.917,6 0,073 -	FV 3 nº1 NOT SIGNIF.	3 n°2 22 148,881 0,000 90,23% - 0,836 0,000	3 n°3 33 42,296 0,000 69,09% 0,257 0,009
Number of observations F statistic P-Value F statistic Adjusted R-squared Significant EV Intercept P-Value Intercept A FV 1 A P-Value \Delta FV 1 A A FV 2 A P-Value \Delta FV 2	FV 3 n°1 10 89.258,102 0,000 87,49% 0,703 0,000 -	3 n°2 NOT SIGNIF	3 n°3 NOT SIGNIF.	88,73% 0,804 0,000 -	3 n°2 22 29,142 0,000 72,83% 4.394.140,6 0,033 0,706	3 n°3 31 32,895 0,000 51,53% 4.860.917,6 0,073 -	FV 3 nº1 NOT SIGNIF.	3 n°2 22 148,881 0,000 90,23% 0,836 0,000 0,480	3 n°3 33 42,296 0,000 69,09% 0,257 0,009 0,140

Table A.3 - Parameters estimation and tests of significance for Model 3.

Model 3 - regr. (ii)		2009			2010			2011	
Sub-sample	WORLD	ITALY	USA	WORLD	ITALY	USA	WORLD	ITALY	USA
Number of observations	48	9	25	NOT SIGNIF.	NOT SIGNIF.	NOT SIGNIF.	65	14	NOT SIGNIF.
F statistic	106,358	8,161	55,111	-	-	-	319,776	107,223	-
P-V alue F statistic	0,000	0,024	0,000	-	-	-	0,000	0,000	-
Adjusted R- squared	67,23%	47,23%	65,50%	-	-	-	81,76%	81,49%	-
Significant EV									
Intercept	-	0,4	-	-	-	-	-	-	-
P-V alue Intercept	-	0,014	-	-	-	-	-	-	-
Δ% FV 2 A	-	1,035	-	-	-	-	5,625	5,705	-
P-Value ∆% FV 2 A	-	0,024	-	-	-	-	0,000	0,000	-
Δ% FV 3 A	2,030	-	2,040	-	-	-	-	-	-
P-Value ∆% FV 3 A	0,000	-	0,000	-	-	-	-	-	-
Sub-sample	PERC. FV 3 n°1	PERC. FV 3 n°2	PERC. FV 3 n°3	PERC. FV 3 n°1	PERC. FV 3 n°2	PERC. FV	PERC.	PERC. FV	PERC. FV
			311 3	1.4 2 11 1	3 n-2	3 n°3	FV 3 n°1	3 n°2	3 n°3
Number of observations	NOT SIGNIF.	NOT SIGNIF.	23	NOT SIGNIF.	NOT SIGNIF.	3 n°3	NOT SIGNIF.	NOT SIGNIF.	3 n°3
		NOT		NOT	NOT		NOT	NOT	
observations		NOT	23	NOT	NOT	31	NOT	NOT	32
observations F statistic P-Value F		NOT	23 412,566	NOT	NOT	31 9,803	NOT	NOT	32 168,700
observations F statistic P-Value F statistic Adjusted R-		NOT	23 412,566 <i>0,000</i>	NOT	NOT	31 9,803 <i>0,004</i>	NOT	NOT	32 168,700 <i>0,000</i>
observations F statistic P-Value F statistic Adjusted R- squared		NOT	23 412,566 <i>0,000</i>	NOT	NOT	31 9,803 <i>0,004</i>	NOT	NOT	32 168,700 <i>0,000</i>
observations F statistic P-Value F statistic Adjusted R- squared Significant EV		NOT	23 412,566 0,000 90,39%	NOT	NOT	31 9,803 0,004 21,29%	NOT	NOT	32 168,700 <i>0,000</i>
observations F statistic P-Value F statistic Adjusted R-squared Significant EV Intercept P-Value Intercept $\Delta\%$ FV 2 A		NOT	23 412,566 0,000 90,39%	NOT	NOT	31 9,803 0,004 21,29%	NOT	NOT	32 168,700 <i>0,000</i>
observations F statistic P-Value F statistic Adjusted R-squared Significant EV Intercept P-Value Intercept		NOT	23 412,566 0,000 90,39%	NOT	NOT	31 9,803 0,004 21,29%	NOT	NOT	32 168,700 <i>0,000</i> 81,25%
observations F statistic P-Value F statistic Adjusted R-squared Significant EV Intercept P-Value Intercept $\Delta\%$ FV 2 A P-Value $\Delta\%$ FV		NOT	23 412,566 0,000 90,39%	NOT	NOT	31 9,803 0,004 21,29%	NOT	NOT SIGNIF. - - -	32 168,700 0,000 81,25%

Table A.4 - Parameters estimation and tests of significance for Model 4.

Model 4 - regr. (ii)		2009			2010			2011	
Sub-sample	WORLD	ITALY	USA	WORLD	ITALY	USA	WORLD	ITALY	USA
Number of	NOT	NOT	NOT	NOT	9	NOT	40	12	NOT
observations	SIGNIF.	SIGNIF.	SIGNIF.	SIGNIF.	,	SIGNIF.	40	12	SIGNIF.
F statistic	-	-	-	-	12,200	-	106,949	439,226	-
P-Value F	_	_	_	_	0,010	_	0,000	0,000	_
statistic					0,0.0		0,000	0,000	
Adjusted R-	-	_	_	-	47,90%	_	81,89%	88,47%	_
squared									
Significant EV									
Intercept	-	-	-	-	-	-	-	-	-
P-Value Intercept	-	-	-	-	-	-	-	-	-
$\Delta\%$ FV 2 net	-	-	-	-	-1,570	-	-4,485	-4,906	-
P-Value $\Delta\%$ FV	-	-	-	-	0,008	-	0,000	0,000	-
2 net $\Delta\%$ FV 3 net			-			-	F4.460		
$\Delta\%$ FV 3 net $P\text{-}V$ alue $\Delta\%$ FV	-	-	-	-	-	-	-54,468	-	-
P-V aine Δ % FV 3 net	-	-	-	-	-	-	0,000	-	-
7 1101	777.0			 			 		
Sub-sample	PERC. FV 3 nº1	PERC. FV 3 n°2	PERC. FV 3 n°3	PERC. FV 3 n°1	PERC. FV 3 n°2	PERC. FV 3 n°3	PERC.		PERC. FV 3 n°3
				1.4 2 11 1	311 4	3 n-3	FV 3 n°1	3 n°2	3 n-3
Number of	NO	NOT		·					311-3
Number of observations	NO Data	NOT SIGNIF	NOT	NO	NOT	NOT	NOT	NO	24
observations	NO Data	NOT SIGNIF.		·					24
observations F statistic			NOT	NO	NOT	NOT	NOT	NO	24 279,707
observations			NOT	NO	NOT	NOT	NOT	NO	24
observations F statistic P-Value F			NOT	NO	NOT	NOT	NOT	NO	24 279,707 <i>0,000</i>
observations F statistic P-Value F statistic			NOT	NO	NOT	NOT	NOT	NO	24 279,707
observations F statistic P-Value F statistic Adjusted R-			NOT	NO	NOT	NOT	NOT	NO	24 279,707 <i>0,000</i>
observations F statistic P-Value F statistic Adjusted R- squared			NOT	NO	NOT	NOT	NOT	NO	24 279,707 <i>0,000</i>
observations F statistic P-Value F statistic Adjusted R- squared Significant EV			NOT SIGNIF. - -	NO	NOT SIGNIF. - -	NOT SIGNIF. - -	NOT	NO DATA - -	24 279,707 <i>0,000</i>
observations F statistic P-Value F statistic Adjusted R- squared Significant EV Intercept			NOT SIGNIF. - -	NO	NOT SIGNIF. - -	NOT SIGNIF. - -	NOT	NO DATA - - -	24 279,707 <i>0,000</i>
observations F statistic P-Value F statistic Adjusted R- squared Significant EV Intercept P-Value Intercept			NOT SIGNIF.	NO	NOT SIGNIF. - -	NOT SIGNIF. - -	NOT	NO DATA - - - -	24 279,707 0,000 91,50%
observations F statistic P-Value F statistic Adjusted R-squared Significant EV Intercept P-Value Intercept $\Delta\%$ FV 2 net			NOT SIGNIF.	NO	NOT SIGNIF. - -	NOT SIGNIF.	NOT	NO DATA - - -	24 279,707 0,000 91,50%
observations F statistic P-Value F statistic Adjusted R-squared Significant EV Intercept P-Value Intercept $\Delta\%$ FV 2 net P-Value $\Delta\%$ FV			NOT SIGNIF.	NO	NOT SIGNIF. - -	NOT SIGNIF. - -	NOT	NO DATA - - - -	24 279,707 0,000 91,50%
observations F statistic P-Value F statistic Adjusted R-squared Significant EV Intercept P-Value Intercept $\Delta\%$ FV 2 net P-Value $\Delta\%$ FV 2 net		SIGNIF.	NOT SIGNIF.	NO DATA - - -	NOT SIGNIF. - - -	NOT SIGNIF.	NOT SIGNIF.	NO DATA - - - -	24 279,707 0,000 91,50% - - -5,101 0,000

Table A.5 – Sample of companies, selected by the criteria specified in section 3.

	Number of	
Country	companies	Currency
Austria	2	Eur
Belgium	1	Eur
Brazil	2	Usd
Canada	1	Usd
Chile	1	Usd
France	12	Eur
Germany	3	Eur
Ireland	2	Eur
Italy	18	Eur
Kenya	1	Usd
Panama	1	Usd
Perù	1	Usd
Puerto Rico	4	Usd
Slovenia	1	Eur
Spain	1	Eur
Sweden	3	Usd
Switzerland	2	Usd
Britain	5	Usd
United States	220	Usd