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A critical validation of the Value Added Intellectual Coefficient: use in empirical research and comparison with alternative measures of intellectual capital

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ABSTRACT

The measurement of intellectual capital (IC) constitutes a major challenge in managing intangible resources. Among the various models proposed in prior literature, the Value Added Intellectual Coefficient (VAIC) is used by many studies to measure IC. Assuming a perfectly competitive market, this study decomposes the VAIC and demonstrates that it is not directly related to IC. Conversely, the main components of VAIC are the labor share, physical capital share, and interest rate. These results are extended to a non-perfectly competitive setting through a multivariate analysis of a cross-country panel of 50,310 firm-year observations for 2000-2017. The results show that the VAIC still largely depends on exogenous factors being negatively (positively) associated with the labor (physical capital) share. Nevertheless, in this non-perfectly competitive setting, the VAIC also captures a firm's ability to generate profits, which may be attributable to multiple factors, including IC. To reduce potential measurement biases in empirical research using the VAIC, this study suggests controlling for a firm's interest rate, labor and capital shares. Adopting this suggestion, this study investigates the association between VAIC and firm performance. The results show that this association is significantly weaker when including the interest rate, labor and capital shares. The theoretical and empirical results suggest future researchers to select the VAIC to measure IC after having conscientiously examined the alternative models proposed in recent literature.

Keywords: VAIC, intellectual capital, value added intellectual coefficient, labor share, capital share

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1. INTRODUCTION

Intellectual capital (IC) is an important topic in management and governance studies. The continuous attention to IC lies in the growing awareness that intangible resources play a central role in the value creation process. The accurate management of IC critically contributes to a firm's wealth (Zambon *et al.*, 2019). Consequently, measuring and reporting IC constitutes a strategic challenge for firms. Although prior studies identify various models to measure IC (e.g., Stewart, 1997; Lev, 2001; 2003; Corrado *et al.*, 2004), managerial literature highlights a critical lack of contributions validating or falsifying these models (Pucci *et al.*, 2015). One of the most frequently used measures of IC is Pulic's (2000) Value Added Intellectual Coefficient (VAIC) (Dumay, 2014; Pedro *et al.*, 2018). This paper adds to the debate related to IC measurement by providing a validation of the VAIC. Theoretical and empirical results show that the VAIC depends largely on exogenous technological factors that are not usually associated with IC. The study also provides potential avenues to overcome the limitations of the VAIC and to assist researchers in the more careful use of the VAIC to measure IC. Finally, theoretical and empirical results are discussed in light of the alternative approaches to measuring and reporting IC proposed in recent literature.

IC and intangible assets have been studied in the literature on marketing and management (Bontis, 1999; Hilmola *et al.*, 2009), organization (Knott *et al.*, 2003), talent management (Sparrow *et al.*, 2014; McCracken *et al.*, 2018), and finance and accounting (Aboody and Lev, 2000; Sullivan and Sullivan, 2000; Basu and Waymire, 2008). In governance studies, prior literature highlights the importance of corporate governance for managing and developing IC (Zambon *et al.*, 2019). Governance mechanisms should be aligned to IC and human capital in particular to contribute to future value creation (Lajili,

2015). Directors' human capital positively influences firm performance (Volontè and Gantenbein, 2016), board turnover after an incident of fraud (D'Onza and Rigolini, 2017), and compensation after IPO (Williams *et al.*, 2018). The level of human capital can also positively influence internationalization (Cerrato and Piva, 2012). Finally, relational capital represented by heterogeneous relational ties among the board of directors has a strong impact on a firm's performance (Rossoni *et al.*, 2018).

Among all intangible assets, IC is typically considered strategic because it relates to human resources, customer relationships, customer loyalty, etc. (Lev, 2005; Pucci *et al.*, 2015). These assets are fundamental both in the internal managerial focus on value creation processes and the external focus on disclosure and reporting on value creation (Badia *et al.*, 2019; Brosnan *et al.*, 2019). This holds true not only in the private sector but also in the public sector, where some universities adopt performance management systems, including IC management, as a criterion for evaluating their managers (Veltri and Puntillo, 2019). Similarly, relational capital fosters collaboration among officers in different ministries as a leading indicator of collaborative governance (Ramadass *et al.*, 2018).

Because IC significantly contributes to value creation, its measurement becomes a primary managerial objective to assess the efficient use and creation of IC (Laing *et al.*, 2010). Badia *et al.* (2019, p. 299) note that “several attempts have been made to develop measurement and reporting systems for this relevant source of value, but managerial literature has highlighted the limits and difficulties of some of those endeavors”. Accounting research tends to consider IC as a part of intangible assets, thus emphasizing its financial dimension and the need for a reliable measurement and contribution to future economic benefits (Skinner, 2008a; 2008b; Guthrie *et al.*, 2012). Conversely, the managerial literature is

interested in understanding how IC contributes to value creation (Johanson *et al.*, 2001; Mouritsen *et al.*, 2001). A more recent approach suggests that IC narratives are strongly relevant, in addition to IC numbers. IC numbers, discourse and disclosure, business models, and financial statements represent a corpus that contributes to the dissemination of the firm's approach to value creation (Mouritsen and Roslender, 2009; Nielsen and Roslender, 2015). While IC numbers cannot stand alone (Gowthorpe, 2009), IC measurement remains a fundamental input to internal managerial decision making (Brosnan *et al.*, 2019) and external firm assessment (Badia *et al.*, 2019).

The VAIC model for measuring IC was initially developed by Pulic (2000) and has rapidly become extremely popular in academic research (Dumay, 2014; Pedro *et al.*, 2018). Prior literature employing the VAIC focuses on IC performance in the banking sector (Mavridis, 2004; Goh, 2005; Kamath, 2007; Alhassan and Asare, 2016), the relationship between IC and market value (Chen *et al.*, 2005), the association between IC and financial performance (Sharabati *et al.*, 2010; Komnenic and Pokrajčić, 2012; Dženopoljac *et al.*, 2016; 2019), and the relationship between organizational capital and cost stickiness (Mohammadi and Taherkhani, 2017). One reason prompting many researchers to select the VAIC as a proxy for IC is that the VAIC intends to measure the capacity of a firm to transform its stock of IC into value added (Pulic, 2008; Iazzolino and Laise, 2013). Additionally, the VAIC measures IC by relying on accounting numbers, which provides researchers with an easy and ready-to-use proxy.

Despite these undeniable advantages, prior studies challenge the validity of the VAIC as a measure of IC. For example, Ståhle *et al.* (2011) argue that the VAIC strongly depends on labor and physical capital investments and thus constitutes a poor measure of IC. In an

attempt to reconcile the debate around the validity of the VAIC, Iazzolino and Laise (2013) suggest that the VAIC is more of a multidimensional measure rather than a direct proxy for IC. While prior literature establishes the limitations of the VAIC as a measure of IC, a systematic theoretical and empirical validation is missing. Consequently, researchers and professionals are left with no guidance regarding the use of the VAIC due to the lack of papers suggesting potential solutions to overcome its inherent limitations. This paper fills this gap in four steps.

First, this study reflects on the critical decomposition of the VAIC provided in Ståhle *et al.* (2011). By assuming a perfectly competitive setting in which profit is equal to zero, this study theoretically demonstrates that the VAIC is a function of labor share, physical capital share, and interest rate. Moreover, the relationship between the VAIC, the labor share, and the physical capital share is nonlinear. These theoretical results imply that the VAIC is not directly related to IC when markets are perfectly competitive. Consequently, they contribute and advance prior knowledge on the critical decomposition of the VAIC.

Second, this study extends the validity of theoretical findings with a multivariate analysis using archival data. In this setting, the hypothesis of perfect competition is relaxed, and the firm's profits are allowed to differ from zero. Under this scenario, the VAIC also captures the ability of a firm to generate profits, which may be attributable to multiple factors, including IC. By using a cross-country dataset of 50,310 firms observed over the 2000-2017 period, the multivariate analysis confirms that the VAIC is negatively (positively) associated with the labor (physical capital) share. The model controls for the time-invariant characteristics for any country, industry, and firm by including firm fixed effects and clustering standard errors at the firm level. Despite the absence of perfect competition, these

empirical results show that the VAIC still depends largely on exogenous technological factors that are not usually associated with IC. Hence, they inform the empirical IC literature on the need to control for technological factors to avoid endogeneity bias in multivariate analysis.

Third, reflecting on theoretical and empirical findings, this study proposes to control for a firm's labor share, capital shares, and interest rate to reduce potential measurement biases in empirical research. The present study adopts this solution to investigate the association between IC, proxied by the VAIC, and financial performance. The results show that the VAIC is positively and significantly associated with financial performance when the labor share, capital share, and the interest rate are not included in the model. To ensure that these results are not driven by the technological components of the VAIC and the interest rate, the model is augmented by controlling for these factors. The results from these tests highlight that the association between the VAIC and financial performance is significantly weaker after controlling for its inherent technological components and the interest rate. As a result, this study prompts a careful interpretation of empirical analysis in prior literature when controls for technological factors are missing.

Fourth, the paper discusses the VAIC in light of the theoretical and empirical validation proposed in this study. The VAIC, which claims to represent IC in numbers only, is compared to alternative models of IC reporting by drawing on a multidimensional and performative approach (e.g., Mouritsen, 2006; Gowthorpe, 2009; Melloni, 2015; Nielsen and Roslender, 2015; Melloni *et al.*, 2016; Bini *et al.*, 2017; Corbella *et al.*, 2019). In empirical research, when properly tuned according to the validation proposed in this paper, the VAIC is capable of serving as a leading indicator of some *potential* relation between IC and the main dependent variable of interest. However, the VAIC does not allow a deeper investigation on

the dimensions of IC (e.g., relational, structural, or human capital) that influence the variable of interest. Similarly, the VAIC cannot explain how IC contributes to value creation. While not discouraging the use of VAIC, this paper suggests empirical literature to select alternative measures of IC that may be more capable of identifying meaningful associations.

The theoretical and empirical findings from this study are important because the VAIC is extensively used in empirical analyses. Researchers are urged to control for the labor share, capital share, and interest rate and to cautiously interpret the results based on the VAIC as a measure of IC. In fact, empirical studies that do not consider these variables as important determinants of the VAIC may suffer from important endogeneity problems. Additionally, the discussion stemming from the results of this paper leads to a reflection on the use of the VAIC in comparison to alternative models of IC reporting.

The remainder of the paper proceeds as follows. Section 2 describes the VAIC model, its underlying assumptions, and results from prior literature adopting this model. Section 3 proposes a decomposition of the VAIC in a perfectly competitive market and theoretically demonstrates that the VAIC is a function of exogenous technological factors that are not usually associated with IC. Section 4 illustrates the methodology and the results of a multivariate analysis that extends the theoretical findings to a non-perfectly competitive setting. In the spirit of assisting researchers in overcoming the VAIC limitations, Section 5 proposes and applies a potential solution for tuning the VAIC to more carefully measure IC. Section 6 offers a discussion of the main findings, outlines the limitations of this study, and provides avenues for future research.

2. REVIEW OF THE LITERATURE ON THE VAIC MODEL

2.1 The VAIC model

The VAIC model intends to offer a methodology to measure intellectual work efficiency and, therefore, IC efficiency (Pulic, 2008; Iazzolino and Laise, 2013). The formula of the VAIC is obtained through the following line of reasoning (Pulic, 2005; Ståhle *et al.*, 2011). First, value added (*VA*) is a function of the structural capital (*SC*) and the human capital (*HC*) employed; *VA* is defined as follows:

$$VA = OP + D + A + C, \quad (1a)$$

where *OP* is the operating profit, *D* is depreciation, *A* is amortization, and *C* is all employee-related costs. According to Equation (1a), *C* identifies the *HC*, while (*OP + D + A*) measures *SC*. Thus, Equation (1a) can also be expressed as follows:

$$VA = SC + HC. \quad (1b)$$

The second step to derive the VAIC formula defines human capital efficiency (*HCE*), structural capital efficiency (*SCE*), and capital employed efficiency (*CEE*) as follows:

$$HCE = VA/HC, \quad (2a)$$

$$SCE = SC/VA, \quad (2b)$$

$$CEE = VA/CE, \quad (2c)$$

where *CE* represents the physical capital employed. It is worth noting that *HCE* focuses on the contribution of human resources to *VA* creation, while *SCE* measures the efficiency of all other components in creating *VA* (Dženopoljac *et al.*, 2016).

Finally, VAIC is the sum of these three efficiency measures:

$$VAIC = HCE + SCE + CEE. \quad (3)$$

Thus, in its aggregated form, the VAIC emphasizes a firm's total efficiency, decomposed into IC efficiency ($ICE = HCE + SCE$) and financial efficiency (CEE).

2.2 Empirical literature using the VAIC model

Pedro *et al.* (2018) document that the VAIC was used in 11% of their reviewed papers, resulting in the most popular proxy for IC.¹ Prior literature initially concentrated on understanding the determinants of VAIC, its variation over time and across countries. Goh (2005) finds that commercial banks in Malaysia show higher human capital efficiency than structural and capital efficiencies. Additionally, they document that domestic banks were generally less efficient than foreign banks and that public banks improve their IC efficiency during the three years covered in their study. These results are confirmed for Indian firms operating in the banking and pharmaceutical industry (Kamath, 2007; 2008), Japanese banks (Mavridis, 2004), and ASEAN countries (i.e., Indonesia, Malaysia, Philippines, Singapore, and Thailand) (Nimtrakoon, 2015). However, firms operating in different segments show significant differences in their VAIC (Mavridis, 2004; Kamath, 2007). Similarly, firms operating in different countries tend to differently emphasize the components of the VAIC (Nimtrakoon, 2015). Finally, El-Bannany (2008) documents that profitability and risk are the main determinants of the VAIC for a sample of UK firms during the 1999-2005 period.

Another strand of research concentrates on analyzing the relationship between the VAIC and firm performance. The results from these empirical studies are mixed. Chen *et al.* (2005) document a positive impact of the VAIC on the market value and financial

¹ More details on the VAIC method and prior empirical literature employing VAIC can be found in Pulic (2000; 2008); Stähle *et al.* (2011); Iazzolino and Laise (2013); Dumay (2014); Pedro *et al.* (2018).

performance of Taiwanese listed firms. They also highlight that the VAIC is a leading indicator of future financial performance. These results are confirmed for financial firms in different countries (Ting and Lean, 2009) and non-financial firms worldwide (Díez *et al.*, 2010; Vishnu and Gupta, 2014; Nimtrakoon, 2015). Additionally, some studies suggest that a specific component of the VAIC has a stronger correlation with financial performance than other components (Chu *et al.*, 2011; Clarke *et al.*, 2011).

Although the above-reported studies point towards a positive relationship between VAIC and financial performance, other studies document inconsistent and, sometimes, opposite results. Some studies support a positive relationship between the VAIC and financial performance for specific industries only (Zeghal and Maaloul, 2010; Pucar, 2012), for human capital only (Madininos *et al.*, 2011; Komnenic and Pokrajčić, 2012; Joshi *et al.*, 2013), or for capital-employed efficiency only (Dženopoljac *et al.*, 2016). Other studies show mixed associations between the components of the VAIC and financial performance (Firer and Williams, 2003; Kamath, 2008; Ghosh and Mondal, 2009; Mehralian *et al.*, 2012). Sardo and Serrasquero (2017) highlight that these mixed results for the components of the VAIC depend on their ability to impact financial performance in the short or long run. Bayraktarouglo *et al.* (2019) identify moderating variables in the relationship between the VAIC and financial performance. Moreover, in their study on Brazilian real estate companies, Britto *et al.* (2014) show a negative relationship between VAIC and both market value and return on invested capital. Similar findings are documented in different countries (Dženopoljac *et al.*, 2017) and in the public sector (Morariu, 2014).

Finally, other studies concentrate on the potential use of the VAIC as a management tool. In their study on the Australian hotel industry, Laing *et al.* (2010) find that managers can

use the VAIC model as a robust tool for assessing the efficient use of IC in their organizations. Additionally, Mohammadi and Taherkhani (2017) document no relation between the VAIC and cost stickiness.

3. THEORETICAL DECOMPOSITION OF THE VAIC IN A PERFECTLY COMPETITIVE SETTING

Despite the widespread use of the VAIC, prior studies challenge its validity as a measure of IC by critically analyzing its constructs. In particular, Stähle *et al.* (2011) suggest that the VAIC depends on labor and physical capital investments. This study elaborates on this intuition and provides a theoretical decomposition of VAIC in a perfectly competitive market.

Because the VAIC computation is based on the relationships between VA and its components, this study first recalls some results related to these relationships in perfectly competitive markets with a standard specification of the production function. In particular, it is assumed that the final output results from three different inputs (human capital, physical capital, and an intermediate good) and a traditional Cobb-Douglas production function, as follows:

$$Y = F(H, K, I) = AH^\alpha K^\beta I^\gamma, \quad \text{with } \alpha, \beta, \gamma > 0, \quad (4)$$

where Y is the total output, A is the state of technology, H is the stock of human capital, K is the stock of physical capital, I is the flow of the intermediate good, and α , β , and γ are the elasticity coefficients of human capital, physical capital, and an intermediate good, respectively. In general, an elasticity coefficient indicates the percentage change in the production level when the corresponding input changes by one percentage point. Given Equation (4), the profit (Π) function is:

$$\Pi = pY - wH - rK - p_I I, \quad (5)$$

where p is the price level of the final output, w is the nominal wage level, r is the interest rate, and p_I is the price of the intermediate input. Because markets are perfectly competitive, the firm is a price-taker in all markets. From Equation (5), the amount of VA the firm generates can also be derived:

$$VA \equiv pY - p_I I = \Pi + wH + rK. \quad (6)$$

Equation (7) provides the extended expression of the VAIC:

$$VAIC \equiv \underbrace{\frac{VA}{wH}}_{HCE} + \underbrace{\frac{VA - wH}{VA}}_{SCE} + \underbrace{\frac{VA}{K}}_{CEE}. \quad (7)$$

Using Equation (6) yields:

$$VAIC = \frac{\Pi + wH + rK}{wH} + 1 - \frac{wH}{\Pi + wH + rK} + \frac{\Pi + wH + rK}{K}. \quad (8)$$

In the long run, the assumption of a perfectly competitive market for the final good implies no profits (i.e., $\Pi=0$). Therefore, Equation (8) becomes:

$$VAIC = 1 + \frac{rK}{wH} + 1 - \frac{wH}{wH + rK} + \frac{wH}{K} + r. \quad (8')$$

Because firms want to maximize their profits, the first-order conditions (FOCs) of Equation (5) must be computed. In particular, the FOCs for H and K lead to the following redistributive results: $wH = \alpha pY$ and $rK = \beta pY$. This result also implies that $\alpha = wH/pY$ and $\beta = rK/pY$, which means that human and physical capital elasticities (α and β) are equal to the labor and capital share, respectively (wH/pY and rK/pY). Moreover, the relation $(wH)/(rK) = \alpha/\beta$ can be written. By inserting these basic results into the operative definition of VAIC, the following equation can be obtained:

$$\text{VAIC} = \underbrace{1 + \frac{\beta}{\alpha}}_{\text{HCE}} + \underbrace{1 - \frac{\alpha}{\alpha + \beta}}_{\text{SCE}} + \underbrace{\frac{\alpha}{\beta}r + r}_{\text{CEE}}. \quad (9)$$

When markets are perfectly competitive and in long-run equilibrium, Equation (9) demonstrates that the VAIC is a function of the elasticity coefficient of human capital (α), the elasticity coefficient of physical capital (β), and the interest rate (r). In particular, both the *HCE* and the *SCE* components negatively depend on α and positively depend on β , while the opposite is true for the *CEE* component. Thus, while the relationship between the VAIC and r is certainly positive, the relationship between the VAIC and the elasticity coefficients is ambiguous.

To further investigate the relationship between the VAIC, the elasticity coefficient of human capital (α), and the elasticity coefficient of physical capital (β), this study numerically simulates Equation (9) by assuming reasonable values for the study parameters (i.e., α and β range between 0 and 1). Figures 1-4 report different VAIC functions for different interest rate levels (1%, 5%, 10% and 20%). According to Figures 1 and 2, the relationship between the VAIC and the elasticity coefficients is increasing in β but decreasing in α . In contrast, a nonlinear relationship between the VAIC and coefficients appears when the interest rate increases (Figures 3 and 4). In particular, Figure 4 shows that the VAIC first decreases with small elasticity coefficient values and then increases.

[INSERT FIGURES 1-2-3-4 ABOUT HERE]

To conclude, this analysis shows that the VAIC may have ambiguous behavior when the elasticity coefficients (especially the elasticity coefficient of human capital) are relatively low. *Ceteris paribus*, efficient firms adopting human capital-intensive technologies could

show lower VAIC levels only as a result of their higher *HCE*. In contrast, the VAIC tends to be higher if a firm pays higher interest rates on physical capital. These relations imply that two firms with the same stock of IC and managerial capabilities may show different VAIC levels due to their technological constraints.

4. EMPIRICAL ANALYSIS OF THE DECOMPOSITION OF VAIC

The decomposition of the VAIC in Equation (9) assumes a perfectly competitive market, which in the long run is inherently characterized by the absence of supernormal profits (i.e., the excess profit above the minimum return necessary to keep an organization in business). Under this assumption, the value added is completely exhausted by the factors of production, meaning that the VAIC captures only technological parameters and the cost of physical capital. However, real markets may violate perfect competition, with firms exhibiting non-zero supernormal profits. In this scenario, the VAIC also captures the ability of a firm to generate profits due to multiple factors, including IC. This section extends the validity of theoretical findings with a multivariate analysis employing archival data. In the empirical analysis, we test the association between the VAIC and firm performance while controlling for the capital share, labor share, and cost of capital.

4.1 Empirical model

This study draws upon previous literature (Goebel, 2015) and regresses the VAIC performance on the interest rate, human and physical capital elasticity.² The following

² For clarity in designing the model, we replace the notations α , β , and r in Equation (9) with *LABOR*, *CAPITAL*, and *R*, respectively.

regression model is estimated (firm subscripts are suppressed):

$$\begin{aligned} VAIC_t = & \theta_0 + \theta_1 LABOR_t + \theta_2 CAPITAL_t + \theta_3 R_t + \theta_4 INTANGIBLES_t + \theta_5 R\&D_t \\ & + \theta_6 LEV_t + \theta_7 PAYMENT_t + \theta_8 SIZE_t + \text{Firm FE} + \text{Year FE} + \varepsilon_t \end{aligned} \quad (10)$$

Consistent with prior literature (e.g., Nimtrakoon, 2015; Dženopoljac *et al.*, 2016; Mohammadi and Taherkhani, 2017), the study measures the dependent variable *VAIC* (i.e., Pulic, 1998, 2000) as the sum of *HCE*, *SCE*, and *CEE*. Further details on the calculation of these variables are provided in Section 2 and in the Appendix. With respect to the main variables of interest, *LABOR* and *CAPITAL* represent measures of labor and capital elasticity, respectively, and are proxied by staff expenses (*LABOR*) and capital expenditures (*CAPITAL*) divided by total sales.³ Similarly, *R* represents the interest rate proxied by the cost of debt capital. To further investigate the relationship between *LABOR*, *CAPITAL*, *R* and *VAIC*, Equation (10) is also estimated by replacing *VAIC* with its components (*HCE*, *SCE*, and *CEE*). The study expects to find a negative association between both *HCE* and *SCE* and *LABOR* and a positive association with *CAPITAL*, while the opposite will be true for the *CEE* component.

Following previous literature, the study also includes some additional firm-level controls that can influence the *VAIC* (Goebel, 2015). Accordingly, this study controls for the level of intangible assets (*INTANGIBLES*), the level of research and development expenses (*R&D*), the leverage ratio (*LEV*), the mean industry-year level of staff expenses (*PAYMENT*),

³ We recall from Section 3 that α is the elasticity coefficient of human capital from Equation (4) and is equal to the labor share (wH/pY). Similarly, β is the elasticity coefficient of the physical capital from Equation (4) and is equal to the capital share (pY/rK). Thus, from now on, we use the terms “labor share” and “capital share” to refer to the elasticity coefficients of intellectual and physical capital.

and the natural logarithm of total sales (*SIZE*). Finally, the study also considers country, industry, and firm time-invariant characteristics and cross-sectional variation by adding firm and year fixed effects and clustering standard errors at the firm level. Continuous variables are winsorized at the 1% and 99% levels, except *SIZE*, which is expressed as a natural logarithm. The Appendix provides additional information about variable definitions and their sources.

4.2 Sample description

The sample includes all firm-year observations available in COMPUSTAT Global⁴ for the 2000-2017 period. The study drops firm-year observations with missing accounting information required to estimate Equation (10). Singleton groups (i.e., groups with only one observation) are removed, as prior literature demonstrates that including these observations can inflate the statistical significance (Correia, 2015). In addition, observations with a negative book value of equity and a negative *VAIC* are removed, as a negative value of “value-added” does not generate a meaningful analysis (see Firer and Williams, 2003; Shiu, 2006; Chan, 2009; Chu et al., 2011; Pal and Soriya, 2012). Consequently, the final sample consists of 50,310 firm-year observations (i.e., 8,401 firms across 81 countries).

Panel A of Table 1 reports the sample distribution by industry-year. Most observations are clustered in the information technology (11,041), industrials (11,120), materials (7,624), consumer discretionary (7,755), and healthcare (5,454) industry sectors. Panel B of Table 1 reveals that most observations are from India (8,815), Taiwan (4,767), the UK (4,519), China (3,253), and Germany (2,994).

[INSERT TABLE 1 ABOUT HERE]

⁴ Accessed via WRDS on September 16, 2018.

4.3 Descriptive statistics and univariate analysis

Table 2 reports the descriptive statistics for the dependent and independent variables entering the multivariate analysis. The mean (median) *VAIC* measure is 7.52 (2.7), while the mean (median) values of *HCE*, *SCE*, and *CEE* are 6.43 (1.81), 0.55 (0.47), and 0.33 (0.26), respectively. Focusing on the main variables of interest, the mean (median) *LABOR*, *CAPITAL* and *R* values are 0.21 (0.15), 0.09 (0.04), and 0.10 (0.06), respectively. On average (median), intangible assets represent 12% (3.4%) of total assets (*INTANGIBLES*). Related to intangibles, research and development expenditures (*R&D*) are approximately 2.9% (1%). Finally, the leverage ratio (*LEV*) is on average (median) 1.74 (1.1), while only 6.2% of firms in the study's sample have a level of staff compensation higher than the mean value by sector-year (*PAYMENT*).

[INSERT TABLE 2 ABOUT HERE]

Table 3 presents the Pearson correlation coefficients for all variables in the multivariate analysis. Consistent with Equation (9), it is observed that *LABOR* is negatively (positively) and significantly correlated with *HCE* (*CEE*) ($p < 0.01$). Notice that, contrary to theoretical predictions, the association between *LABOR* and *SCE* is positive and statistically significant ($p < 0.01$). However, this coefficient may be misleading because it is based on simple, univariate correlations. This fact leads to a negative and significant association between *LABOR* and *VAIC* ($p < 0.01$). Fully consistent with the prediction from Equation (9), *CAPITAL* is negatively correlated with *CEE* ($p < 0.01$) and positively correlated with *HCE* ($p < 0.01$) and *SCE* ($p < 0.01$), resulting in a positive and significant association between *CAPITAL* and *VAIC* ($p < 0.01$).

[INSERT TABLE 3 ABOUT HERE]

4.4 Multivariate analysis

Table 4 reports the results for the tests regarding the influence of interest rate, labor and capital shares on the VAIC. Column (1) shows the baseline model that includes labor share (*LABOR*), capital share (*CAPITAL*), and the interest rate (*R*) without including firm and year fixed effects. Columns (2) and (3) progressively augment the baseline model with the controls from Equation (10). The results confirm that the theoretical model based on perfectly competitive markets also remains valid in a non-perfectly competitive setting. In fact, Table 4 shows that the labor share (*LABOR*) has a negative and statistically significant association with *VAIC* ($p < 0.01$), while the capital share (*CAPITAL*) has a positive and statistically significant association with *VAIC* ($p < 0.01$). However, the relation between the interest rate (*R*) and *VAIC* is not statistically significant. As shown in the theoretical derivation, this result occurs because the interest rate enters only the CEE component of *VAIC*. All other controls are in line with prior literature (e.g., Goebel, 2015). The multivariate analysis confirms that, despite not being directly related to a firm's IC, labor and capital shares are two fundamental drivers of *VAIC*.

[INSERT TABLE 4 ABOUT HERE]

Table 5 reports the results for the investigation into the influence of *LABOR*, *CAPITAL*, and *R* on the *VAIC* components. Accordingly, *VAIC* is replaced with *HCE*, *SCE*, and *CEE*, and the model is estimated in Equation (10). The results show that *CAPITAL* is positively correlated with *HCE* and *SCE*, negatively correlated with *CEE*, and statistically significant at a conventional level ($p < 0.01$). Table 5 also shows that *LABOR* is negatively

correlated with *HCE* ($p < 0.01$) and positively correlated *CEE* ($p < 0.05$). Finally, *LABOR* is not linearly correlated with *SCE*. This lack of correlation may be caused by the linearization of a nonlinear expression. Additionally, *R* is positively correlated with *SCE* and *CEE*. These findings are in line with our theoretical results and confirm that *LABOR*, *CAPITAL* and *R* are important drivers of *VAIC* and its components (*HCE*, *SCE*, and *CEE*) in a non-perfectly competitive setting.

[INSERT TABLE 5 ABOUT HERE]

5. TUNING THE VAIC

The preceding two sections demonstrate that the *VAIC* is largely influenced by exogenous technological factors that are unrelated to *IC*. While this result constitutes a criticism of the use of the *VAIC* as a measure of *IC*, in a non-perfect competitive setting, the *VAIC* also captures the ability of a firm to generate profits. This ability may be attributable to numerous circumstances, including *IC*. In the spirit of assisting empirical researchers, this section shows a potential solution for tuning the *VAIC* to more carefully measure *IC*.

Stemming from prior section's results, this paper argues that controlling for the interest rate, labor and capital share in an empirical analysis is fundamental to avoiding potential biases. Not including *LABOR*, *CAPITAL*, and *R* as additional controls may lead to serious omitted variable concerns and biased estimations. By contrast, including *LABOR*, *CAPITAL*, and *R* helps in ensuring that *VAIC* does not proxy for exogenous technological factors and the interest rate. Far from perfectly proxying for *IC*, *VAIC* will then capture the ability of a firm to generate profits and, therefore, time-varying heterogeneity across firms.

To test the potential solution identified above, this study investigates the association between VAIC and financial performance. Drawing upon previous literature (Pal and Soriya, 2012; Sardo and Serrasqueiro, 2017; Dženopoljac *et al.*, 2019), the following regression model is estimated (firm subscripts are suppressed):

$$\begin{aligned}
 PERFORMANCE_t = & \theta_0 + \theta_1 VAIC_t + \theta_2 LABOR_t + \theta_3 CAPITAL_t + \theta_4 R_t \\
 & + \theta_5 INTANGIBLES_t + \theta_6 R\&D_t + \theta_7 LEV_t + \theta_8 PAYMENT_t + \theta_9 SIZE_t \\
 & + \text{Firm FE} + \text{Year FE} + \varepsilon_t
 \end{aligned} \tag{11}$$

where *PERFORMANCE* is return on assets (*ROA*), return on equity (*ROE*), and return on investments (*ROI*), alternatively, and all other variables are as defined earlier. Equation (11) is estimated excluding *LABOR*, *CAPITAL*, and *R*, then excluding *VAIC* and, finally, including all these variables. If not controlling for *LABOR*, *CAPITAL*, and *R* generates biased estimations, the coefficient of *VAIC* obtained when non-controlling for these variables should be significantly different than that when controlling for these factors.

Table 6 reports the results of the estimation of Equation (11) with alternative measures of financial performance. The results show that the *VAIC* is significantly associated with financial performance (0.018, Column (1); 0.051, Column (4); 0.079, Column (7); $p < 0.01$). Similarly, *LABOR*, *CAPITAL*, and *R* are also associated with financial performance (-6.178, 1.749, and 1.175 Column (2); -13.597, 5.427, and 2.027, Column (5); -30.592, 8.743, and 3.884, Column (8)). Finally, when *VAIC*, *LABOR*, *CAPITAL*, and *R* are included in the estimation of Equation (11), the results show that all these variables are significantly related to financial performance. However, when controlling for *LABOR*, *CAPITAL*, and *R*, the

coefficient of *VAIC* is significantly different than when not controlling for these factors.⁵ Thus, the estimations produced, including *LABOR*, *CAPITAL*, and *R*, are more careful and ensure that the association between the *VAIC* and financial performance is not overestimated. When controlling for technological factors, explanatory power varies from 45.7% when using ROE as a measure of financial performance to 66.5% when employing ROI (adjusted *r*-squared in columns (6) and (9), respectively). This difference can be attributed to the underlying difference in calculating ROE, ROA, and ROI. In fact, ROE and ROA include non-operating items that are not considered when calculating ROI. As *VAIC* refers to a firm's operations, the explanatory power is higher when using ROI to proxy financial performance than when using ROA and ROE.

[INSERT TABLE 6 ABOUT HERE]

6. DISCUSSION AND CONCLUSIONS

The use of the *VAIC* as a measure of IC is consistently growing due to its reliance on accounting numbers and its ease of calculation. Motivated by prior critiques of the *VAIC*'s validity, this study offers a theoretical decomposition of the *VAIC* in a perfectly competitive setting and demonstrates that the *VAIC* is a function of the elasticity coefficient of human capital, the elasticity coefficient of physical capital, and the interest rate. The theoretical results are extended to non-perfectly competitive settings through a multivariate analysis of a cross-country panel of 50,310 firm-year observations covering the years 2000-2017. Reflecting on the theoretical and empirical findings, this study proposes guidance for the use

⁵ A Wald test shows that the difference between the estimated coefficients of the *VAIC* in Column (1) and Column (3) is statistically significant at $p < 0.01$, that between those in Column (4) and Column (6) is statistically significant at $p < 0.10$, and that between those in Column (7) and Column (9) is statistically significant at $p < 0.01$.

of the VAIC to overcome its inherent limitations. More specifically, this study suggests that future empirical literature control for a firm's interest rate, labor and capital shares to reduce potential measurement biases. In an additional analysis investigating the association between the VAIC and financial performance, the results show that this association is weaker when controlling for the interest rate, labor and capital share, prompting a more careful interpretation of the results.

The results of this study prompt a reflection on the use of the VAIC as a measure of IC and should be interpreted in light of the growing body of literature suggesting a multidimensional and performative approach to IC. This paper suggests that the VAIC construct is unrelated to IC from a theoretical perspective that assumes a perfectly competitive market. Conversely, from an empirical perspective, it shows that when relaxing the perfect competition assumption, the VAIC captures the ability of a firm to generate profits. Hence, the VAIC can be *cautiously* used as a leading indicator of IC. When properly tuned according to the validation proposed in this paper, academic studies may adopt the VAIC in empirical analyses to identify some *potential* relation between IC and the main dependent variable of interest. In case they find a significant relation, they are advised to investigate more deeply in order to identify the component of IC and the mechanisms through which IC affects the main dependent variable of interest.

Conversely, the VAIC can provide little or no information on how IC contributes to the value creation process. In fact, the concept of IC has many dark sides, which cannot be easily captured in the VAIC numbers or other models (Gowthorpe, 2009). As a result, the correlation between IC numbers is sometimes useless (Mouritsen and Roslender, 2009). Hence, recent literature clearly outlines that the metrics of IC cannot stand alone but should

be read and interpreted in connection to disclosure following a performative approach (Mouritsen, 2006).

Although there is no single solution to measuring and reporting IC, prior literature identifies some practice-based patterns. For example, Corbella *et al.* (2019) demonstrate that some concepts inspired by the International Integrated Reporting Council Framework are considered pivotal in defining categories of IC and identifying its contribution to value creation. Nielsen and Roslender (2015) show how business model discourse and disclosure can contribute to improving the information content of financial statements by also including information on IC and its contribution to value creation. Another strand of research concentrates on the interaction between business models and non-financial key performance indicators to obtain a more integrated framework that can provide more relevant and performative disclosure to users (Bini *et al.*, 2017; 2018). While empirical content analysis demonstrates a tendency toward impression management (Melloni, 2015; Melloni *et al.*, 2016), integrated reporting, business models, key performance indicators, and IC definitely show a strong link and influence each other in practice (Zambon *et al.*, 2019).

Compared to the VAIC, these approaches are certainly more exhaustive in explaining the contribution of IC to the value creation process. Despite its inherent limitations, the VAIC offers a ready-to-use measure based on accounting numbers. Conversely, collecting and analyzing information conveyed through integrated reporting, business models, and key performance indicators is certainly more complicated and costlier. In the end, the two approaches serve two different purposes. The VAIC provides a superficial tool for measuring IC that can be used to identify some potential relation to value creation while being incapable of any deeper investigation. Conversely, integrated reporting, business models, and key

performance indicators allow a granular – albeit time and resource consuming – understanding of the contribution of IC to the value creation process.

Despite its contributions, this study is subject to some potential limitations that generate avenues for future research. First, this study limits its theoretical decomposition of the VAIC to a perfectly competitive setting. Future studies may question the decomposition of VAIC in different settings, such as monopolies, oligopolies, and other market forms. Second, the study empirically validates its theoretical model with a linear regression model. Future contributions could investigate the existence of a nonlinear relationship between the VAIC, labor share, and capital share. Finally, future research could use nonmonetary measures of market competition, such as the *Herfindahl-Hirschman* index or other concentration indices, to test whether these measures moderate the role of VAIC in explaining firms' performance.

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APPENDIX – VARIABLE DEFINITIONS

Variables	Description (COMPUSTAT GLOBAL code)
VAIC ^w	Sum of HCE, SCE, and CEE.
HCE ^w	Human Capital Efficiency: VA divided by total staff expenses (XLR).
SCE ^w	Structural Capital Efficiency: VA minus staff expenses divided by VA (XLR).
CEE ^w	Capital Employed Efficiency: VA divided by the difference between total assets and the value of intangibles (AT, INTAN).
LABOR ^w	Labor share: Total staff expenses divided by total sales (XLR, SALE).
CAPITAL ^w	Capital share: Total capital expenditures divided by total sales (CAPX, SALE).
R ^w	Interest and related expenses divided by the sum of long-term debt and debt in current liabilities (XINT, DLTT, DLC).
INTANGIBLES ^w	Total intangibles divided by total assets (INTAN, AT).
R&D ^w	Research and development expenses divided by total assets (XRD, AT).
LEV ^w	Total liabilities divided by total common shareholder equity (LT, CEQ).
PAYMENT	Dummy variable that takes the value of 1 if the staff expenses are above the mean value by sector year, 0 otherwise (XLR).
SIZE ^L	Natural logarithm of total assets (AT).
ROA ^w	Income before extraordinary items divided by total assets (IB, AT).
ROE ^w	Income before extraordinary items divided by common shareholder equity (IB, CEQ).
ROI ^w	Earnings before interest and taxes divided by the difference between total assets and current assets (EBIT, AT, ACT).

^w winsorized the 1% and 99% levels, while ^L is the natural logarithm.

TABLES

Table 1: Sample distribution.

Panel A: Sample distribution by Global Industry Classification Sector (GIC) and year.

GIC Sector	2000	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010	2011	2012	2013	2014	2015	2016	2017	Total
10 - Energy	10	18	18	30	36	37	51	70	72	79	84	83	88	91	105	104	93	77	1,146
15 - Materials	155	168	197	211	238	235	322	414	436	486	494	508	510	609	706	715	690	530	7,624
20 - Industrials	224	253	276	287	332	252	422	617	661	714	753	750	769	870	989	1066	1026	859	11,120
25 - Consumer discretionary	110	151	178	179	195	208	274	411	484	481	511	551	573	671	727	750	737	564	7,755
30 - Consumer staples	69	79	109	120	136	132	180	243	265	292	284	287	299	302	326	357	345	278	4,103
35 - Health care	120	144	165	180	203	165	233	326	348	361	358	366	373	394	452	455	448	363	5,454
40 - Financials	6	6	10	17	18	7	10	19	18	20	19	19	14	15	14	12	10	7	241
45 - Information technology	171	201	226	275	284	198	333	531	573	596	622	638	647	1,051	1,208	1,234	1,211	1,042	11,041
50 - Telecom services	10	14	17	21	25	18	26	37	41	43	43	50	49	48	45	42	38	35	602
55 - Utilities	22	28	32	32	41	40	43	73	75	84	74	84	76	83	89	97	97	85	1,155
60 - Real estate	2	2	3	2	3	2	3	4	5	3	1	4	2	6	6	7	8	6	69
Total	899	1,064	1,231	1,354	1,511	1,294	1,897	2,745	2,978	3,159	3,243	3,340	3,400	4,140	4,667	4,839	4,703	3,846	50,310

Panel B: Sample distribution by country and year.

COUNTRY	2000	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010	2011	2012	2013	2014	2015	2016	2017	Total
ARGENTINA	0	0	0	0	0	0	0	2	2	1	2	1	3	5	6	6	5	6	39
AUSTRALIA	4	6	13	18	28	31	91	113	141	135	140	136	132	140	141	141	121	109	1,640
AUSTRIA	9	7	5	8	12	8	20	34	36	30	31	37	34	37	39	40	36	33	456
BANGLADESH	0	0	0	0	0	1	3	7	12	10	13	18	17	20	20	15	15	26	177
BELGIUM	10	9	13	16	18	10	23	39	38	41	41	40	41	41	39	39	38	36	532
BERMUDA	15	32	38	54	49	31	47	82	87	81	83	79	83	70	77	70	60	50	1,088
BRAZIL	1	0	2	2	2	9	11	30	32	43	43	50	56	57	57	50	48	45	538
BULGARIA	0	0	0	0	1	2	2	2	1	1	2	2	0	1	2	3	3	1	23
CAYMAN ISLANDS	8	17	29	38	50	47	88	128	182	199	211	230	250	267	270	287	261	243	2,805
CHILE	0	0	0	0	0	0	0	0	1	7	8	8	9	9	10	12	10	10	84
CHINA	2	15	27	51	49	35	37	49	51	54	57	97	130	149	501	660	661	628	3,253
COLOMBIA	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	2	4	3	10
CROATIA	1	1	2	2	2	3	3	2	4	4	3	3	3	3	2	3	2	2	46
CURAÇAO	0	0	1	1	1	1	1	1	2	1	1	1	1	1	0	1	1	2	17
CYPRUS	0	0	0	0	1	1	1	3	4	2	4	3	3	3	4	5	4	3	41
CZECH REPUBLIC	0	1	1	2	3	3	1	3	4	3	3	3	2	3	2	2	2	1	39
DENMARK	18	18	25	23	19	18	24	39	37	39	40	44	40	45	44	41	39	35	588
EGYPT	0	0	0	0	0	0	0	0	0	0	1	3	4	3	5	7	4	4	31
ESTONIA	0	0	0	0	0	0	2	4	3	2	2	1	1	1	2	2	1	1	22
FAROE ISLANDS	0	0	0	0	0	0	0	0	1	1	0	1	1	1	1	1	1	1	9
FINLAND	47	56	56	48	53	32	58	77	76	72	76	71	72	73	72	76	71	65	1,151
FRANCE	61	70	80	93	103	72	90	140	152	173	180	197	186	185	198	195	184	157	2,516
GERMANY	61	77	88	107	120	50	98	213	226	225	236	234	222	227	212	208	203	187	2,994
GREECE	1	5	11	15	23	25	33	30	39	43	41	37	37	32	30	31	23	22	478
HONG KONG	6	6	9	12	10	9	14	14	20	20	22	27	29	31	35	38	41	33	376
HUNGARY	0	0	1	0	1	2	2	3	3	1	2	2	2	4	3	6	5	5	42
ICELAND	2	2	2	2	2	2	2	4	2	2	2	2	2	2	2	2	2	2	38
INDIA	211	260	290	287	330	435	525	592	585	627	622	570	538	537	537	579	565	95	8,185
INDONESIA	5	4	5	8	2	11	11	19	19	22	24	23	22	27	31	30	33	28	324
IRELAND	10	12	10	8	8	4	11	12	13	15	13	10	11	10	10	13	13	12	195
ISLE OF MAN	0	0	0	0	0	0	0	0	0	0	1	1	0	2	3	3	3	2	15
ISRAEL	13	10	16	17	21	22	47	55	69	82	79	79	77	77	78	69	75	61	947
ITALY	1	3	5	15	38	23	28	50	66	71	78	80	76	80	77	78	85	68	922
JERSEY	3	3	2	3	2	1	1	3	5	5	7	9	13	11	11	11	7	4	101
JORDAN	1	1	2	1	2	1	2	0	3	1	1	5	3	5	6	8	9	10	61
KAZAKHSTAN	0	1	1	1	0	1	1	1	1	1	1	1	1	1	1	1	1	1	16
KOREA REP.	0	3	1	4	5	4	3	0	0	0	0	0	1	1	1	0	0	0	24
LATVIA	2	1	1	0	1	2	1	3	4	5	5	5	6	5	3	5	5	5	59
LITHUANIA	0	0	0	0	0	0	0	1	3	3	2	3	3	2	2	2	2	2	25
LUXEMBOURG	1	2	3	1	3	1	4	6	7	10	13	12	13	15	13	13	14	10	141
MALAWI	0	0	0	0	0	0	0	0	0	0	0	1	0	1	1	1	1	1	6
MALAYSIA	12	19	36	39	44	55	57	122	125	130	131	136	126	129	125	116	100	80	1,582

MALTA	0	0	0	0	0	1	1	1	2	3	4	2	4	2	2	2	2	2	28
MAURITIUS	1	1	0	0	0	0	0	0	0	0	0	0	1	1	1	1	0	0	6
MEXICO	0	0	0	0	0	0	1	3	1	2	1	2	6	6	6	8	9	8	53
MOROCCO	0	0	0	0	0	1	1	1	0	1	0	1	2	0	0	3	2	1	13
NETHERLANDS	14	14	17	18	24	18	36	43	44	48	47	48	50	49	46	41	36	34	627
NEW ZEALAND	0	0	1	1	1	0	5	13	20	23	24	29	26	30	22	18	21	22	256
NIGERIA	1	1	2	4	3	2	3	3	2	3	2	1	2	2	2	1	5	5	44
NORWAY	11	12	14	17	21	11	27	43	39	46	46	39	39	33	38	36	35	34	541
OMAN	0	0	0	0	0	1	2	2	1	1	1	0	0	1	1	2	1	0	13
PAKISTAN	9	12	7	5	3	6	9	22	31	28	21	24	21	18	17	16	17	16	282
PALESTINE	0	0	0	0	0	0	0	0	0	0	1	1	2	1	1	1	0	0	7
PERU	0	0	0	0	0	0	0	0	0	1	1	2	2	4	4	4	4	4	26
PHILIPPINES	1	4	5	5	7	10	9	20	23	25	26	30	27	25	22	24	24	26	313
POLAND	3	2	3	3	6	8	7	9	17	31	46	63	83	72	72	77	74	64	640
PORTUGAL	0	0	0	0	0	0	2	1	3	5	6	6	7	5	5	5	6	7	58
ROMANIA	0	0	0	1	1	3	1	2	3	4	6	4	4	6	8	5	4	5	57
RUSSIAN FEDERATION	1	2	3	3	5	6	11	18	17	20	19	17	21	25	23	24	22	14	251
SAUDI ARABIA	0	0	0	1	2	1	1	3	3	4	6	8	7	7	7	9	9	9	77
SERBIA	0	0	0	0	0	0	1	2	3	2	2	3	2	2	1	1	1	1	21
SINGAPORE	11	19	27	33	40	25	35	73	79	78	64	61	52	61	64	64	62	49	897
SLOVAKIA	1	1	0	0	1	1	0	1	1	1	1	0	1	2	0	1	2	1	15
SLOVENIA	0	0	0	0	0	1	3	3	2	3	7	6	4	5	4	2	2	2	44
SOUTH AFRICA	13	17	25	25	27	13	33	44	50	54	48	49	48	52	43	46	43	37	667
SPAIN	1	1	3	5	3	9	14	20	22	27	36	43	48	53	53	55	54	49	496
SRI LANKA	2	1	0	0	1	1	5	9	9	10	11	11	12	11	15	17	14	9	138
SWEDEN	48	52	54	67	69	52	33	78	84	104	97	99	96	82	79	88	94	83	1,359
SWITZERLAND	49	43	52	47	46	22	62	83	81	83	91	84	91	81	78	73	73	72	1,211
TAIWAN	0	1	2	3	8	9	6	7	12	13	35	57	69	783	963	958	953	888	4,767
THAILAND	0	0	0	0	1	1	0	3	4	5	11	10	15	14	15	17	18	13	127
TRINIDAD AND TOBAGO	0	0	0	0	0	0	1	1	0	0	0	1	1	1	1	1	1	1	9
TUNISIA	0	0	0	0	0	0	1	2	2	2	1	2	4	4	4	2	4	2	30
TURKEY	1	2	2	5	6	27	32	43	52	55	50	74	94	100	98	101	98	95	935
UGANDA	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	1	0	0	2
UKRAINE	0	0	0	0	0	0	0	0	0	0	0	0	1	2	3	1	1	1	9
UNITED ARAB EMIRATES	0	0	0	0	0	1	1	1	0	0	0	1	1	1	1	2	2	2	13
UNITED KINGDOM	225	236	237	232	228	108	208	302	309	304	301	291	296	275	284	249	241	193	4,519
VIET NAM	0	0	0	0	0	0	0	0	0	0	1	0	1	0	1	4	4	5	16
VIRGIN ISLANDS	1	1	1	2	3	2	3	5	4	8	8	8	9	10	7	6	5	3	86
ZIMBABWE	1	1	1	1	2	2	1	1	2	2	1	1	1	1	1	1	1	0	21
<i>TOTAL</i>	<i>899</i>	<i>1,064</i>	<i>1,231</i>	<i>1,354</i>	<i>1,511</i>	<i>1,294</i>	<i>1,897</i>	<i>2,745</i>	<i>2,978</i>	<i>3,159</i>	<i>3,243</i>	<i>3,340</i>	<i>3,400</i>	<i>4,140</i>	<i>4,667</i>	<i>4,839</i>	<i>4,703</i>	<i>3,846</i>	<i>50,310</i>

Table 2: Descriptive statistics.

Variable	N. Obs.	Mean	St. Dev.	Min	25°	Median	75°	Max
<i>VAIC</i>	50,310	7.527	19.675	0.001	2.098	2.700	4.081	133.759
<i>HCE</i>	50,310	6.435	18.763	-1.590	1.352	1.806	3.024	125.887
<i>SCE</i>	50,310	0.548	0.609	-0.994	0.283	0.470	0.702	5.391
<i>CEE</i>	50,310	0.326	0.282	-0.915	0.151	0.256	0.424	1.665
<i>LABOR</i>	50,310	0.215	0.465	0.001	0.070	0.151	0.260	9.924
<i>CAPITAL</i>	50,310	0.086	0.244	0.000	0.019	0.040	0.084	6.362
<i>R</i>	50,310	0.101	0.242	0.000	0.033	0.056	0.088	2.318
<i>INTANGIBLES</i>	50,310	0.116	0.166	0.000	0.003	0.034	0.166	0.703
<i>R&D</i>	50,310	0.029	0.057	0.000	0.002	0.010	0.031	0.546
<i>LEV</i>	50,310	1.741	2.580	0.006	0.602	1.096	1.938	26.522
<i>PAYMENT</i>	50,310	0.062	0.242	0	0	0	0	1
<i>SIZE</i>	50,310	7.183	2.485	-1.041	5.452	7.214	8.750	26.890

See the Appendix for variable definitions.

Table 3: Pearson correlation coefficients.

		1	2	3	4	5	6	7	8	9	10	11	12
<i>VAIC</i>	1	1											
<i>HCE</i>	2	0.988**	1										
<i>SCE</i>	3	0.234**	0.168**	1									
<i>CEE</i>	4	-0.142**	-0.145**	-0.344**	1								
<i>LABOR</i>	5	-0.108**	-0.126**	0.175**	0.067**	1							
<i>CAPITAL</i>	6	0.021**	0.017**	0.180**	-0.159**	0.432**	1						
<i>R</i>	7	-0.011*	-0.015**	0.037**	0.063**	0.046**	-0.023**	1					
<i>INTANGIBLES</i>	8	0.025**	0.019**	-0.040**	0.514**	0.094**	-0.057**	0.010*	1				
<i>R&D</i>	9	-0.031**	-0.041**	0.101**	0.155**	0.305**	0.023**	0.075**	0.092**	1			
<i>LEV</i>	10	0.002	0.002	-0.016**	-0.032**	-0.003	-0.009*	-0.043**	-0.003	-0.033**	1		
<i>PAYMENT</i>	11	-0.051**	-0.050**	-0.027**	0.012**	-0.011*	0.006	-0.017**	-0.022**	-0.010*	0.030**	1	
<i>SIZE</i>	12	0.106**	0.116**	0.025**	-0.248**	-0.229**	0.020**	-0.104**	-0.128**	-0.279**	0.055**	0.460**	1

Number of observations: 50,310. See the Appendix for variable definitions. **, and * denote significance at the 1% and 5% levels, respectively.

Table 4: Association between labor share (*LABOR*), capital share (*CAPITAL*), interest rate (*R*), and VAIC.

Dependent variable	Column (1) VAIC	Column (2) VAIC	Column (3) VAIC
Constant	8.271*** (43.84)	8.648*** (41.81)	-0.972 (-0.54)
LABOR	-6.068*** (-9.90)	-6.278*** (-5.79)	-5.862*** (-5.44)
CAPITAL	6.715*** (8.67)	2.770*** (3.89)	2.446*** (3.49)
R	-0.177 (-0.61)	-0.098 (-0.27)	0.036 (0.10)
INTANGIBLES			-1.341 (-0.80)
R&D			-2.606 (-1.07)
LEV			0.108** (1.97)
PAYMENT			-12.469*** (-9.25)
SIZE			1.444*** (5.75)
Year FE	No	Yes	Yes
Firm FE	No	Yes	Yes
Observations	50,310	50,310	50,310
Adjusted R^2	0.017	0.363	0.371
F	36.514	11.389	15.851

See the Appendix for variable definitions. t statistics are in parentheses and based on clustered standard errors at the firm level. ***, **, and * denote significance at the 1%, 5%, and 10% levels, respectively.

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Table 5: Association between labor share (*LABOR*), capital share (*CAPITAL*), interest rate (*R*), and the components of VAIC.

Dependent variable(s):	Column (1) HCE	Column (2) SCE	Column (3) CEE
Constant	-1.903 (-1.11)	0.495*** (9.67)	0.540*** (22.63)
LABOR	-6.423*** (-7.54)	-0.003 (-0.09)	0.029** (2.48)
CAPITAL	2.559*** (4.06)	0.135*** (3.38)	-0.047*** (-5.96)
R	-0.137 (-0.41)	0.036** (2.06)	0.015*** (2.84)
INTANGIBLES	-1.428 (-0.89)	-0.064 (-1.07)	0.435*** (16.86)
R&D	-2.145 (-0.96)	0.596*** (2.88)	0.000 (0.00)
LEV	0.087* (1.68)	0.003 (1.45)	-0.003*** (-5.24)
PAYMENT	-12.176*** (-9.53)	-0.110*** (-9.09)	0.069*** (10.58)
SIZE	1.441*** (6.04)	0.004 (0.57)	-0.037*** (-11.09)
Year FE	Yes	Yes	Yes
Firm FE	Yes	Yes	Yes
Observations	50,310	50,310	50,310
Adjusted R^2	0.381	0.480	0.821
F	20.461	15.198	64.891

See the Appendix for variable definitions. t statistics are in parentheses and based on clustered standard errors at the firm level. ***, **, and * denote significance at the 1%, 5%, and 10% levels, respectively.

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Table 6: Association between labor share (*LABOR*), capital share (*CAPITAL*), interest rate (*R*), VAIC and firm performance.

Dependent variable(s):	Column (1) ROA	Column (2) ROA	Column (3) ROA	Column (4) ROE	Column (5) ROE	Column (6) ROE	Column (7) ROI	Column (8) ROI	Column (9) ROI
Constant	7.053*** (5.49)	8.643*** (6.88)	8.652*** (6.89)	22.390*** (5.83)	25.968*** (6.80)	26.000*** (6.82)	39.338*** (6.48)	47.762*** (8.02)	47.800*** (8.03)
VAIC	0.018*** (8.03)		0.010*** (3.67)	0.051*** (5.80)		0.034*** (3.54)	0.079*** (7.53)		0.040*** (2.93)
LABOR		-6.178*** (-7.87)	-6.120*** (-7.76)		-13.597*** (-6.87)	-13.401*** (-6.75)		-30.592*** (-6.17)	-30.359*** (-6.09)
CAPITAL		1.749** (2.44)	1.725** (2.42)		5.427*** (2.85)	5.345*** (2.81)		8.743** (2.30)	8.645** (2.28)
R		1.175*** (2.99)	1.175*** (2.99)		2.027** (2.15)	2.026** (2.14)		3.884** (2.08)	3.882** (2.08)
INTANGIBLES	-5.389*** (-3.90)	-5.491*** (-4.23)	-5.478*** (-4.22)	-8.917** (-2.11)	-9.143** (-2.25)	-9.098** (-2.24)	-17.464** (-2.20)	-18.240** (-2.37)	-18.187** (-2.37)
R&D	-74.400*** (-11.10)	-72.430*** (-11.06)	-72.405*** (-11.05)	-172.330*** (-10.75)	-168.037*** (-10.57)	-167.950*** (-10.57)	-232.512*** (-6.95)	-222.952*** (-6.95)	-222.849*** (-6.95)
LEV	-0.727*** (-14.46)	-0.716*** (-14.14)	-0.717*** (-14.16)	-5.835*** (-14.73)	-5.811*** (-14.67)	-5.814*** (-14.69)	-1.318*** (-7.39)	-1.274*** (-7.17)	-1.279*** (-7.19)
PAYMENT	-0.242 (-0.78)	0.038 (0.13)	0.161 (0.55)	0.237 (0.23)	0.725 (0.73)	1.143 (1.14)	-0.116 (-0.10)	1.404 (1.20)	1.900 (1.62)
SIZE	0.014 (0.08)	-0.052 (-0.30)	-0.067 (-0.39)	-0.194 (-0.36)	-0.350 (-0.66)	-0.398 (-0.75)	-1.708** (-2.07)	-2.092*** (-2.60)	-2.150*** (-2.67)
Year FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Firm FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	50,310	50,310	50,310	50,310	50,310	50,310	50,310	50,310	50,310
Adjusted R^2	0.556	0.570	0.570	0.450	0.457	0.457	0.652	0.665	0.665
F	71.243	54.341	56.987	66.551	52.534	51.880	30.580	23.477	27.564
β VAIC 1 vs.3; 4 vs.6; 7 vs. 9	-	-	8.85***	-	-	3.41*	-	-	8.59***

See the Appendix for variable definitions. t statistics are in parentheses and based on clustered standard errors at the firm level. ***, **, and * denote significance at the 1%, 5%, and 10% levels, respectively. We multiplied the dependent variable by 100 for ease of representation. On the basis of a Wald test, the last row shows that the difference

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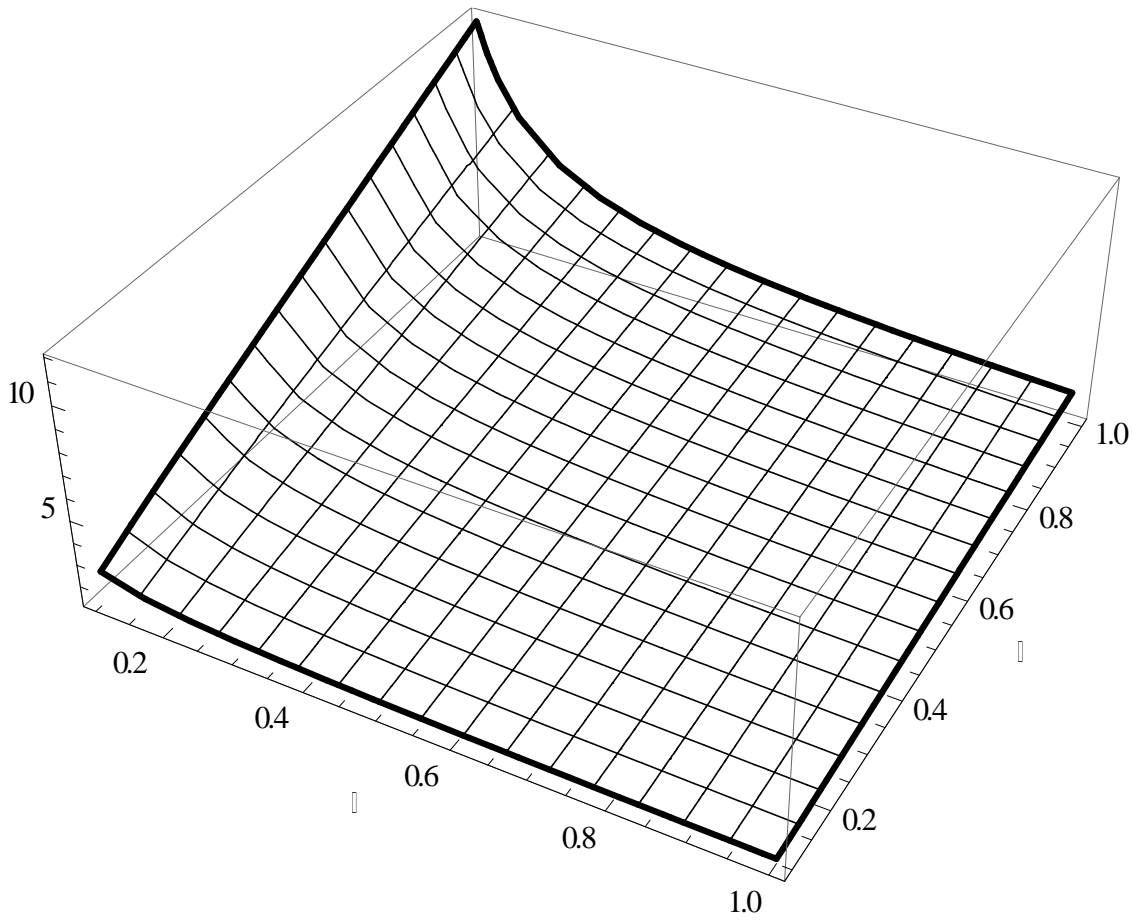
between the estimated coefficients of VAIC in Column (1) and Column (3) is statistically significant at $p < 0.01$ ($F = 8.85$), between those and Column (4) and Column (6) at $p < 0.10$ ($F = 3.41$), and between those in Column (6) and Column (9) at $p < 0.01$ ($F = 8.59$).

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FIGURES

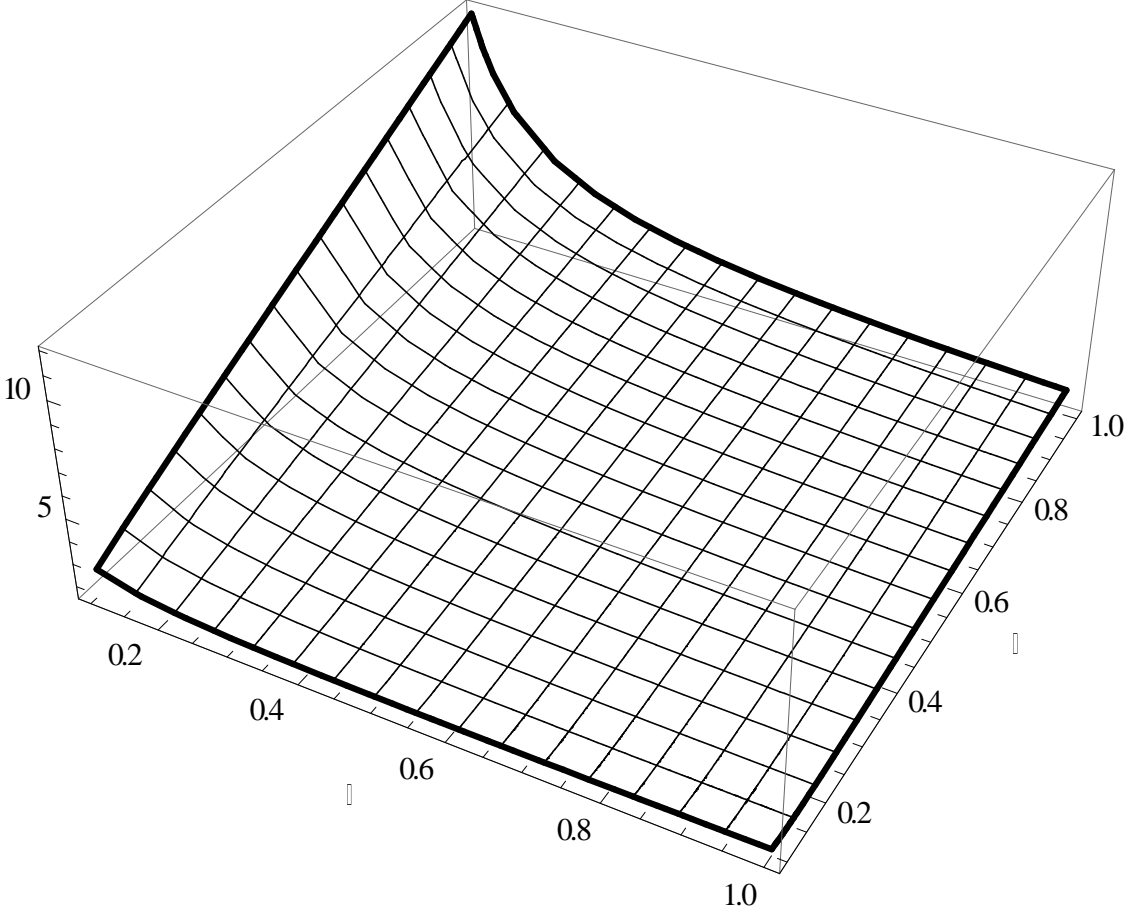
Figure 1: VAIC function ($r=1\%$).



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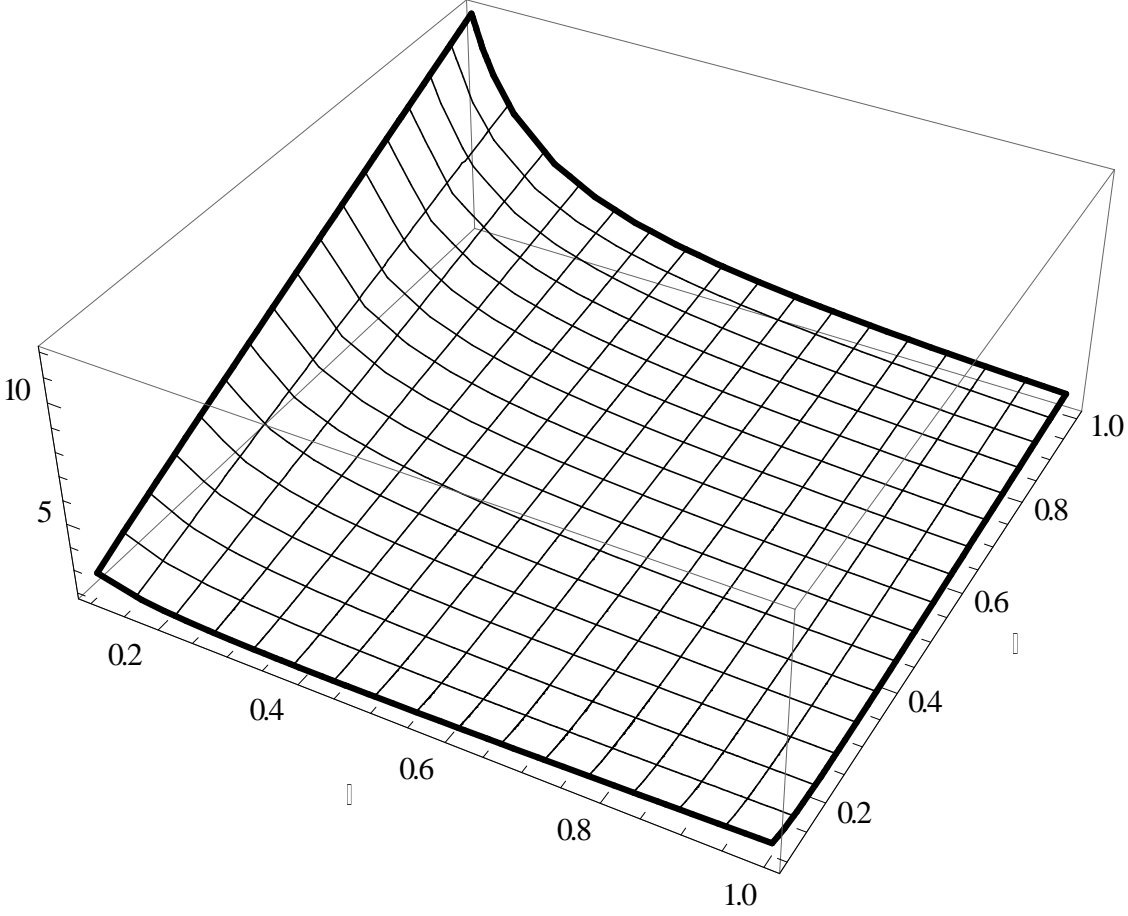
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Figure 2: VAIC function ($r=5\%$).



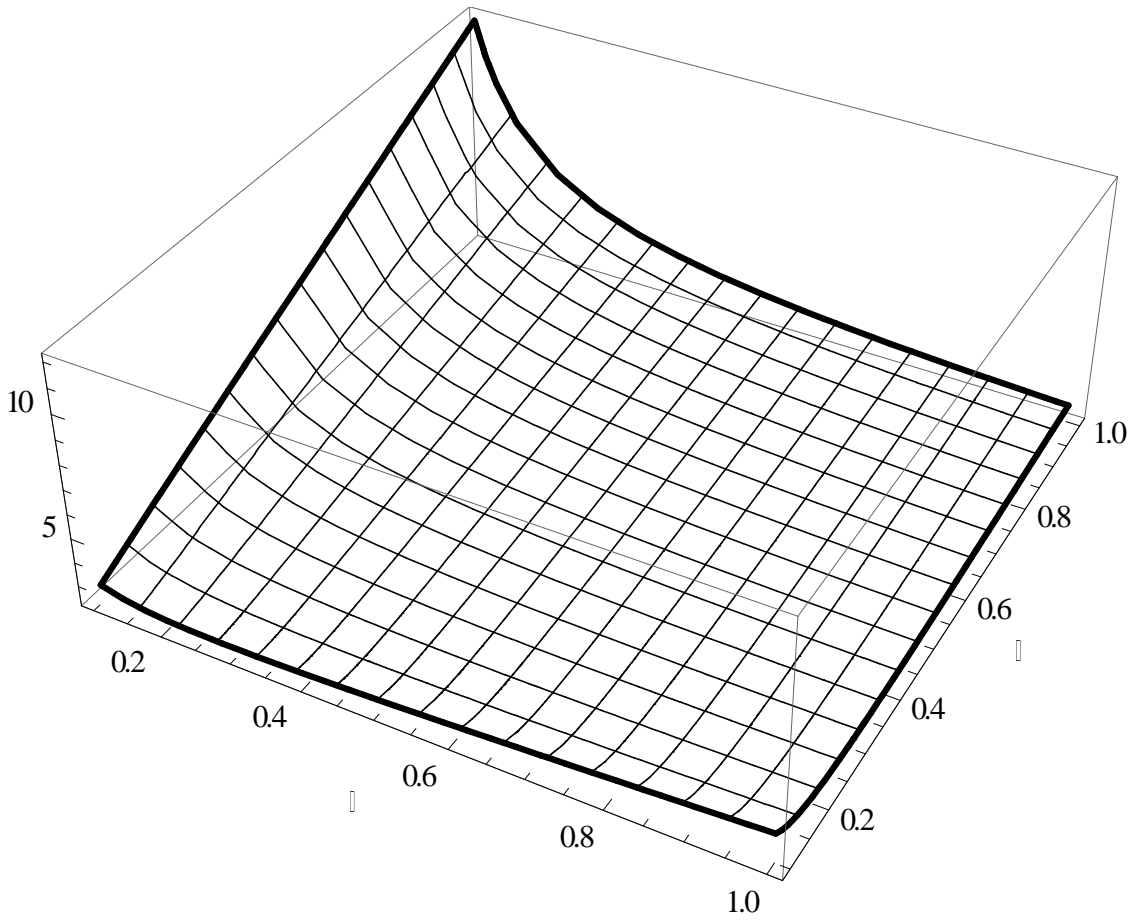
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Figure 3: VAIC function ($r=10\%$).



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Figure 4: VAIC function ($r=20\%$).



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