

Financialisation and intangible assets in emerging market economies: evidence from Brazil

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Whereas previously scholars advocated a positive relationship between a growing size of the financial sector and economic growth, most recent evidence has shown that this might not be the case at all times. The financialisation literature has pointed to some of the mechanisms through which the increasing size and changing structure of the financial system might weigh negatively on growth through the changing financial relations of non-financial corporations (NFCs). This paper contributes to this debate on several grounds. First, rather than interrogating the relationship between finance and firms' tangible investments, it focuses on firms' intangible investments, arguably a sine-qua-non for innovativeness and productivity-enhancing structural change. Drawing on an emerging literature on intangible assets, innovation, and development studies, we highlight the important role of investment into *intangible assets*, in the context of developing economies. Second, by bringing together the literatures on access to finance, intangible assets, and financialisation, we delineate analytically *three specific channels* through which finance can affect intangible assets. Third, this is the first paper that tests empirically all three channels using the population of publicly listed manufacturing companies in an Emerging Market Economy, Brazil over the period 2011–2016. Our results confirm the potentially negative impact of financialisation on intangible assets through the *crowding-out channel*, that is, firm's increased tendency to hold financial assets reduces intangible assets. Our findings also confirm the *shareholder-value orientation channel*, that is, firm's payments of dividends reduce intangibles assets.

Key words: Intangible assets, Financialisation, Innovation-driven growth, Emerging Market Economies

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1. Introduction

The size and complexity of financial markets have increased substantially over recent decades, a process frequently summarised as ‘financialisation’. Although emanating from developed economies, developing and emerging countries (DECs) have not been excluded from this secular change in the workings of the global economy: global capital flows to DECs have surged and domestic financial markets have continued to grow, frequently becoming more market based (Bonizzi, 2013; World Bank, 2013; Unctad, 2019). At the same time though, productive, high-value-added structural change and industrialisation—for many a cornerstone of economic development—has stalled in many DECs regions, in particular Latin America and Africa (Unctad, 2003, 2019; Castillo and Martins Neto, 2016; Rodrik, 2016; Haraguchi *et al.*, 2017).

Traditionally, financial and industrial developments were seen to go hand in hand as financial institutions (banks) provide the necessary savings to finance private investment, diversify risk, and exercise corporate control (Gerschenkron, 1962; King and Levine, 1993a, 1993b; Levine, 1997). Moreover, it was assumed that, as financial systems mature, financial markets, rather than arms-length bank lending, would be more effective in fulfilling these functions (Levine and Zervos, 1996, 1998). Developments since the financial crisis of 2008 inspired contributions that have empirically challenged elements of these views. Studies have shown that the relationship between measures of financial development and economic growth is non-monotonic, and that beyond certain thresholds economies may experience financial hypertrophy (Cecchetti and Kharroubi, 2012; Law and Singh, 2014; Arcand *et al.*, 2015). Others have argued that market-based financial systems may not always be better than bank-based ones (e.g. Beck *et al.*, 2014). This more critical view of finance is also reflected in the interdisciplinary literature on financialisation, which has shown that growing financial markets might weigh on investment, shorten the horizon of managers, and increase payments to financial markets at the expense of long-term investment (Stockhammer, 2004; Orhangazi, 2008; Van Treeck, 2008; Tori and Onaran, 2017).

This paper contributes to this crucial policy question by revisiting the link between financial development and economic growth through the financial operations of non-financial corporation (NFCs) in the context of DECs. It makes three contributions to the literature. First, in contrast to existing studies, which have predominantly focused on firms’ tangible capital investment, it highlights the important nexus between those financial operations and firms’ intangible investment, which boost innovativeness and productivity—arguably one of the most important prerequisites for productive structural change (Marrocu *et al.*, 2012; Bontempi and Mairesse, 2015; Montresor and Vezzani, 2016; Añón Higón *et al.*, 2017). Building on recent developments in the innovation and finance literatures, it focuses on firms’ investments into *intangible assets*. Intangible assets are defined as a portfolio of diverse knowledge activities spanning from technological inputs (e.g. R&D, IT/software) to non-technological inputs (e.g. design, advertising) which, in turn, stimulate the firm’s innovativeness and productivity (Eisfeldt and Papanikolaou, 2014; Bontempi and Mairesse, 2015; Peters and Taylor, 2017). Intangible assets have recently been conceptualised as an important antecedent of firms’ innovativeness and productivity capabilities (Teece, 2000; Montresor and Vezzani, 2016) and leapfrogging in DECs (e.g. Lall, 1992; Dutrénit, 2000). This broad conceptualisation of innovativeness is consistent with how innovation occurs in DECs, which in this context is more about adapting foreign

We test these hypotheses using the population of publicly listed manufacturing companies in Brazil over the period 2011–2016. We apply the [Arellano and Bond \(1991\)](#) two-step difference generalised methods of moments (GMM) estimator. Our empirical strategy accounts for dynamic panel bias, firm fixed effects, time-related shocks, endogenous variables, autocorrelated disturbances and general forms of heteroscedasticity.¹

Our results confirm the potentially negative relationship between financialisation and investments into intangibles through two channels: first, the *crowding-out channel*, that is, firm's increased tendency to hold financial assets reduces intangible assets, and second, the *shareholder-value orientation channel*, that is, firm's shareholder value orientation manifested as high payments of dividends reduces intangibles assets. We find the first effect to be stronger, potentially confirming the higher importance of the crowding-out channel for DECAs, which are characterised by structurally higher interest rates and greater macroeconomic uncertainty than developed countries (see also [Demir 2008, 2009b](#)).

In the next section, we discuss the literature on the differential effects of access to finance and financialisation upon intangible assets. [Section 3](#) elaborates the methodological approach and econometric estimation whereas [Section 4](#) presents the empirical results. Finally, [Section 5](#) provides the concluding remarks and contribution of the study.

2. Finance, financialisation and intangible assets nexus: three channels

2.1 Channel 1: access to finance and intangible assets

Intangible assets include both technological inputs (e.g. R&D, IT/software) and non-technological inputs (e.g. design, advertising). Early research in the field of the economics of innovation paid particular attention to technological forms of intangibles and in particular to investments on R&D or patents ([Griliches, 1979](#)). Recently attention is drawn to non-technological forms of intangibles such as advertising as it is emphasised that they facilitate the commercialisation of inventions (i.e. innovation) and they increase the appropriability of investments on R&D ([Marrocu *et al.*, 2012](#); [Añón Higón *et al.*, 2017](#)). Essentially, both technological and non-technological forms of intangibles work in tandem. They complement each other during the innovation process—from idea generation and creativity to R&D, invention, design, development, and commercialisation—strengthen firm's capabilities, improve manufacturing firm performance ([Añón Higón *et al.*, 2017](#)), and stimulate innovation-driven growth ([Lev, 2001](#)). The increasing importance of intangibles for firm performance in today's global economy has also been highlighted in a recent literature, which shows that it is the use of intangibles (e.g. branding, modern marketing techniques, the access to databases and related AI technologies), rather than tangible capital assets, which allows lead firms in Global Value Chains to enhance their market power ([Durand and Milberg, 2020](#)).

Prior literature has shown that access to finance plays an important role in stimulating investments in intangible assets and inducing innovation ([King and Levine, 1993a, 1993b](#); [Brown *et al.*, 2009](#); [Hsu *et al.*, 2014](#); [Nanda and Kerr, 2015](#); [Pellegrino and Savona, 2017](#)). The [Modigliani–Miller \(1959\)](#) theorem in corporate finance states that returns to different types of investment are independent of the way a corporation

¹ Our results are also robust to alternative model specifications.

3.3 Operationalising key variables

3.3.1 Intangible assets The International Financial Reporting Standards (IFRS), set out the criteria for identifying and measuring intangible assets. Companies are required to disclose this information in their financial accounts. The IFRS defines intangible assets as an identifiable⁷ non-monetary asset without physical substance. They specify the intangible assets that companies should include when measuring their investments on intangibles, these are: ‘Identifiable intangible assets include patents, copyrights, licences, customer lists, brand names, import quotas, computer software, marketing rights and specialised know-how’⁸ (Wiley IFRS, 2017, p. 201). Recent research points out that ‘investment in intangible capital, which includes R&D and the software component of ICT, is largely investment in innovation’ (Corrado *et al.*, 2012; p. 4; Teece, 1986).

Prior studies on innovation have solely focused on measuring innovation via measures such as patents, technological investments such as R&D, or via self-reported surveys on binary outputs of whether a company innovates or not (e.g. Community Innovation Survey).⁹ Although research based on these proxies generated valuable insights, there are certain caveats associated with such measures, for example, not all patents are turned into commercially viable innovations; not all firms invest on formal R&D, especially in DECs; and there is an increased bias of self-reported surveys. The advantage of adopting a measure of innovativeness via a company’s accounts on investments in intangible assets is that it captures not only technological activities such as R&D and patents, but a broader range of activities that contributes to innovation such as ICT, advertising, design, copyrights, brand names etc. A recent report that examines the returns on investment on science and innovation acknowledges that such investments are not limited to R&D but also include a range of ‘intangible investments’ (BIS, 2015). This expansive understanding of innovativeness aligns with the way innovation unfolds in DECs. It primarily involves the adaptation of foreign technologies into local production systems, wherein local firms upgrade in global value chains through the integration of complementary non-technological inputs (Lall, 1992; Kesidou and Romijn, 2008; Bonizzi *et al.*; 2023). This contrasts with the sole focus of research in advanced economies on scientific breakthroughs that are typically associated with patents or R&D.

The importance of intangibles for productivity growth has been initially recognised via the work of Corrado *et al.* (2005, 2009, 2012, 2013). Using a Solow-Jorgenson-Griliches growth model, Corrado and Hulten (2010) show that intangible capital has surpassed tangibles as the largest source of growth. The analysis of these studies was largely at the aggregate level of country or sector. Recent progress in research on intangibles in both innovation and finance literatures, focuses on the micro level and uses company accounts to measure intangibles at the firm level (Arrighetti *et al.*, 2014;

⁷ Note that in order to identify an intangible asset such asset needs to be separable, or to arise from contractual or other legal rights. This is because separable assets can be sold, transferred, licensed, etc. (Wiley IFRS, 2017).

⁸ Goodwill is not included in the financial accounts that measure investments on intangible assets. This is because external goodwill, acquired in a business combination (e.g. during a merger), is outside the scope of investments on intangible assets. On the other hand, internally generated goodwill is within the scope of intangible assets but is not recognised as an asset because it is not an identifiable resource (Wiley IFRS, 2017).

⁹ Patents and R&D are not available in our sample. For robustness, we collected manually patent data and re-run our analysis. The results remain robust, confirming H2.

Eisfeldt and Papanikolaou, 2014; Bontempi and Mairesse, 2015; Montessoro and Vezzani, 2016; Peters and Taylor, 2017).

In sum, investments on intangible assets have been increasingly acknowledged of strategic importance for innovation and productivity in both innovation and finance literatures as it captures the commitment of the company to innovation and organic growth (Corrado *et al.*, 2012). Here, we consider that intangible assets¹⁰ reflect the breadth and depth of a firm’s technological and non-technological innovativeness. We calculate the logarithm of intangibles assets (in millions of US\$).

3.3.2 *Access to finance and financialisation* Table 1 summarises the finance and financialisation indicators we use and their relation to the theoretical mechanisms set out in Section 2. All variables are measured in millions of US \$. To test the effect of access to external finance, that is, *H1*, we measure external financing as firms’ Financial Liabilities (FL), namely firms’ total short-term and long-term debt, as a percentage of firms’ Total Assets (TA). This measure is denoted (FL/TA). A positive impact of this proxy upon intangibles will confer support to *H1*.

To test the effect of financialisation on intangible assets via the *crowding-out* channel, that is, *H2*, we use two proxies: First, Financial Assets (FA) as a percentage of Total Assets (TA), denoted (FA/TA). Higher values of this proxy indicate that companies favour financial investments. Second, Financial Profits (FP) as a percentage of Total Profits (TP), denoted (FP/TP). Higher values of this variable imply that companies make more revenues from financial channels rather than from their underlying innovation or operational activities. We expect both measures to exert a negative influence on intangibles assets.

To test the effect of financialisation on intangibles assets via the *shareholder-value orientation* channel, that is, *H3*, we use two proxies: First, Dividend Payments as a percentage of Total Equity, denoted (Dividend/Equity). Second, Stock repurchases as a percentage of Total Equity, denoted (Stock repurchase/Equity). We expect a negative effect for both proxies, as higher dividend payments and expenditure for the repurchase of equity reduce resources available for the continued accumulation on intangibles assets.

3.4 Model specification and estimation

3.4.1 *Empirical models* We investigate *H1*, *H2* and *H3* by estimating the following model.

$$\begin{aligned} \log(\text{intangibles})_{i,t} = & \beta_0 + \beta_1(\text{FL/TA})_{i,t-1} + \beta_2(\text{FA/TA})_{i,t-1} + \\ & \beta_3(\text{FP/TA})_{i,t-1} + \beta_4(\text{Dividends/Equity})_{i,t-1} + \beta_5(\text{Repurchase/Equity})_{i,t-1} + \\ & \beta_6(\text{Interest payments/TA})_{i,t-1} + \beta_7 \log(\text{TA})_{i,t-1} + d_t + \varepsilon_{i,t} \end{aligned} \quad (1)$$

where the dependent variable is the logarithm of firms’ intangible assets. (FL/TA) is financial liabilities as a percentage of total assets; this variable captures the *access to finance* hypothesis (*H1*). (FA/TA) is financial assets as a percentage of total assets and (FP/TP) is financial profits as a percentage of total profits; both variables capture the *crowding-out* channel of financialisation (*H2*). *Dividends/Equity* is dividend payments

¹⁰ We also conduct robustness analysis whereby we use an alternative dependent variable, intangible intensity, that is, the ratio of intangible assets over total assets. Our results remain fairly consistent.

Table 1. *Finance, financialisation and intangible assets nexus: concepts and measurement*

	Underlying conceptual mechanism	Measurement	Indicator
Dependent Variable Intangible assets		Intangibles assets: we measure a firm's total intangibles assets at cost, not at fair market value (intangibles). This is the total expenditure on assets including patents, copyrights, licences, customer lists, brand names, import quotas, computer software, marketing rights and specialised know-how (excluding goodwill). We divide intangibles with total assets (TA).	Intangible assets: $\log(\text{intangibles})$
Hypothesis 1. Firms with higher access to external finance invest more on intangible assets .	<i>Access to finance channel:</i> firms that are able to borrow externally would also invest more on intangible assets.	External finance: we measure external financing as firms' financial liabilities (FL). This is the total short-term and long-term debt. Further we divide FL with total assets (TA).	External finance: (FL/TA)
Hypothesis 2. Highly financialised firms that allocate a high proportion of their assets on financial assets invest less on intangible assets .	<i>Crowding-out channel:</i> firms make portfolio decision between investing in financial assets or intangible assets. An increase in financial assets and/or financial profits will crowd out investments on intangible assets. We consider that this is manifested via an asset allocation mechanism (via financial assets) and/or a return on investment mechanism (net financial profits).	Financial assets: we measure financial assets as the sum of current assets, including cash and cash equivalents, short-term account receivables, and all other short-term investments excluding goodwill and intangibles (FA). We divide FA with total assets (TA). Financial profit: we measure net financial profits (before taxes) as the sum of dividend income from subsidiaries, interest income and gains from other financial investments, net of all expenses associated with such operations (FP). We divide FP with total profit (before taxes). Total profits is the sum of pre-tax net operating profit, financial profit and net equity (TP).	Financial assets: (FA/TA) Financial profit: (FP/TP)

Table 1. Continued

	Underlying conceptual mechanism	Measurement	Indicator
Hypothesis 3. Highly financialised firms that are exposed to pressures from shareholders invest less on intangible assets .	<i>Shareholder value orientation channel:</i> pressures from shareholders change management behaviour. They strengthen value extraction over value creation, increase dividend and financial payments, and cause stock buybacks. As a result, investments on intangible assets drop.	Dividend payments: we measure dividend payments (dividends) as a percentage of total equity (equity). Stock repurchases: we measure stock repurchases (repurchase) as a percentage of total equity (equity).	Dividend payments: (dividends/equity) Stock repurchases: (repurchase/equity)

Source: Authors.

as a percentage of total equity, *Repurchase/Equity* is firms' repurchase of equity as a percentage of total equity; these variables capture the *shareholder value orientation* channel of financialisation (*H3*).

We control for the size of the firm using the logarithm of firms' total assets (*TA*). We also control for the firm's interest payments as a percentage of total assets (*Interest Payments/TA*), since higher interest repayments may lead to lower real investments and is potentially correlated with payments for dividends and for the repurchase of equity. We include a vector of time dummies, d_t , to account for time-specific unobserved common shocks, such as economic recessions, changes in regulation affecting the Brazilian manufacturing sector, financial crises, and other macroeconomic shocks. We use one-year lags of all the variables because the accumulation of intangibles assets may only respond to changes in companies' asset and income structure with a delay.¹¹ $\varepsilon_{i,t}$ is the error term.

3.4.2 Estimation methods To estimate (1), we employ two methods: a Fixed Effects estimator, and a Difference GMM estimator.¹² We first estimate the following fixed effects model:

$$y_{i,t} = \beta_0 + \beta X_{i,t-1} + d_t + \alpha_i + \varepsilon_{i,t}, \quad (2)$$

where $y_{i,t}$ is the logarithm of intangible assets, $X_{i,t-1}$ is a vector of lagged explanatory variables in equation (1), d_t is a vector of time dummies, $\varepsilon_{i,t}$ is the error

¹¹ In unreported regressions, we estimate the effects of both contemporaneous and lagged values of the variables. We find that financial assets have a negative contemporaneous and lagged effect, but dividends do not. Hence, crowding out may be instantaneous but shareholder orientation takes some time.

¹² We also first explore a simple pooled ordinary least squares (POLS) estimator. The results from this model indicate that access to finance as well as all measures of financialisation exert a negative and statistically significant effect on intangibles assets. The POLS model thus shows no support for *H1*, whilst it supports *H2* and *H3*. Given that this model does not account for firm-specific effects, we use fixed effects and GMM models as our main specifications.

term, α_i represent firm dummies that capture time invariant firm effects, and all other variables are as previously defined. Incorporating firm fixed effects is important because it allows us to account for endogeneity resulting from unobservable, time-invariant factors that may affect both the accumulation of intangible assets and the financialisation behaviour of the firm, such as its location or its organisational culture. This fixed effects model, however, does not account for other sources of endogeneity that are independent of fixed effects, such as reverse causality and simultaneity between the dependent and explanatory variables. Indeed, intangibles assets, financial assets, financial profits and financial liabilities may be jointly determined; firms' investment decisions are not independent and are often made with similar performance-related outcomes in mind. Moreover, investing in or profiting from one asset may encourage or reduce investments in other assets, raising the potential for reverse causality. Even though we rule out contemporaneous reverse causality by using one-year lags of the explanatory variables, this only makes them predetermined with respect to intangibles assets, but they are not independent of past realisations of intangibles assets: they are not strictly exogenous (Roodman, 2009). In addition, the fixed effects model is unsuitable for a dynamic panel specification, which allows current realisations of intangibles assets to be influenced by past ones.¹³ This is important because innovation is a cumulative and persistent process, such that past levels of intangibles would reinforce future levels.

To address these concerns, we employ the Arellano and Bond (1991) Difference GMM estimator.¹⁴ This estimator allows consistent estimation of a dynamic panel model where the lagged dependent variable is included in the model. The Difference GMM estimator first transforms the regression equation in (1) into first differences (Roodman, 2009). The transformed variables are then instrumented with their past levels:

$$\Delta y_{i,t} = \theta \Delta y_{i,t-1} + \beta \Delta X_{i,t-1} + d_t + \Delta \varepsilon_{i,t}, \quad (3)$$

where $\Delta y_{i,t-1}$ is the first differenced one-year lagged intangibles assets (i.e. growth in intangibles), $\Delta X_{i,t-1}$ is the vector of first differenced explanatory variables in equation (1), d_t is a vector of time dummies and $\Delta \varepsilon_{i,t}$ is the differenced error term. Since all the right-hand side variables in (4) are predetermined, then in the absence of second-order serial correlation, the past levels of these variables should only be correlated with the errors dated $t - j$, $j \geq 2$, and not with current disturbances (Roodman, 2009; Baum, 2013). At the same time, these past levels should be strongly correlated with the first differenced variables in (4), so they make good instrumental variables. The validity of this identification strategy depends crucially on the absence of second-order serial correlation and on the exogeneity of the instruments. We test the former using the Arellano–Bond test for second-order serial correlation, and the latter using Hansen J statistic of overriding restrictions.

¹³ In a dynamic specification, the fixed effects estimator is biased because the lagged dependent variable will be correlated with the current error term, introducing a new source of endogeneity. Without a large time dimension to average out the effects of this correlation, the endogeneity problem persists. Kiviet (1995) finds that the bias is 20% of the coefficient even when $T = 30$.

¹⁴ We prefer the difference GMM to the system GMM approach, because the latter requires many more instruments and, with our relatively small sample size, this results in an instrument proliferation issue (see Roodman, 2009).

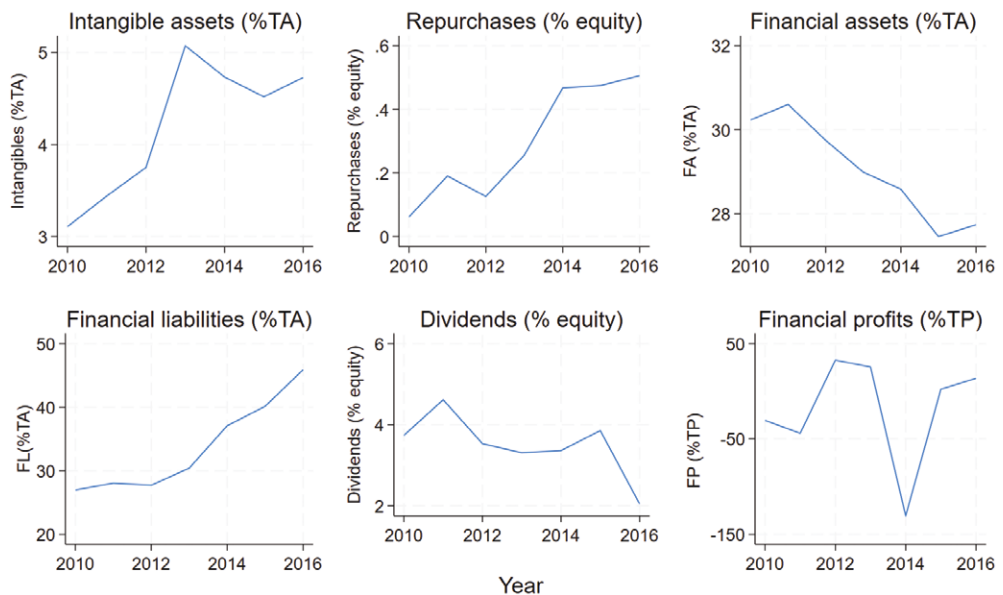
et al., 2022) reports that intangibles are 4.8% of total assets among UK firms. Soener (2021) shows that financial assets (including cash, short-term investments, total current receivables, other financial assets)/total assets for 37 economies that were ever among the largest 40 economies globally since 1988 started at around 30% in 1991 and declined to around 28% in 2017. These are similar values to ours. With regards to firm liabilities as a share of total assets, Davydov (2016) shows that the mean book value of the total debt to total assets ratio for Brazil, Russia, China and India between 2003 and 2012 was 65%. This is higher than our value, which could be explained by the inclusion of China and Russia and the different time period under consideration.

Table 3 shows the pairwise correlation coefficients between the variables. There appears to be little correlation between the various financialisation measures.

Table 2. Descriptive statistics

Variable	N	Mean	SD	Min	Max
log(intangibles)	410	1.93	3.47	-8.09	8.81
FL/TA	410	0.36	0.51	0.00	4.59
FA/TA	410	0.29	0.15	0.00	0.81
FP/TP	410	-0.11	7.29	-96.23	26.35
Dividends/equity	407	-3.44	5.99	-63.33	0.00
Repurchase/equity	402	-0.004	0.02	-0.33	0.00
Interest payments/TA	402	-0.14	0.15	-1.81	0.00
log(TA)	410	6.29	2.03	0.39	10.35

Source: Authors calculations using balance sheet data for Brazilian manufacturing firms.



Notes: TA refers to Total Assets, TO refers to Total Profits.

Fig. 1. Evolution of intangibles assets and financialisation measures over time, (2010–2016)

Source: Authors calculations using balance sheet data for Brazilian manufacturing firms.

Table 3. Pairwise correlation coefficients

Variables	1	2	3	4	5	6	7	8
1 log(intangibles)	1							
2 FL/TA	-0.15*	1						
3 FA/TA	0.05	-0.05	1					
4 FP/TP	-0.04	0.01	0.03	1				
5 Dividends/equity	-0.26*	0.15	-0.21*	0.02	1			
6 Repurchase/equity	-0.08	0.01	-0.09	0.00	0.06	1		
7 log(TA)	0.84*	-0.19*	-0.06	-0.02	-0.23*	-0.06	1	
8 Interest payments/TA	-0.12	-0.08	-0.05	0.02	-0.03	0.11*	-0.02	1

Source: Authors calculations using balance sheet data for Brazilian manufacturing firms.

* Significance at the 1% or 5% level.

4. Results and discussion

The first column of Table 4 shows results from the Fixed Effects model denoted in equation (2). Here, firm's access to external finance, that is, financial liabilities, has a positive impact on intangibles assets, conferring support for *H1*, that is, *access to finance*. Specifically, a 1% increase in financial liabilities relative to total assets increases intangibles assets by 0.5%. The fixed effects model also shows strong support for *H2*, that is, the *crowding-out hypothesis*. Specifically, a 1% increase in financial assets relative to total assets reduces intangibles assets by 4.1%, and a 1% increase in financial profits relative to total profits reduces intangibles assets by 0.09%. The fixed effects model also supports *H3*, that is, the *shareholder value orientation hypothesis*. A 1% increase in dividends paid relative to total equity reduces intangibles assets by 1.3%. Similarly, a 1% increase in repurchase relative to total equity reduces intangibles assets by 1.9%. Taken together, the fixed effects model supports *H1*, *H2* and *H3*.

Next, we consider results from GMM estimates that allow for a dynamic panel specification, enabling us to account for previous levels of intangibles assets. The second and third of Table 4 show the results from short-run and long-run GMM estimates implied by equation (3). The coefficient of the lagged dependent variable is positive and statistically significant. This supports the view that the accumulation of intangibles assets is a cumulative and persistent process, so that previous levels of intangibles assets reinforce future intangibles. Here, firms' access to external finance, that is, financial liabilities have a positive but insignificant impact in both the short-run and long-run. In contrast to the fixed effects model, this implies no support for *H1*. The GMM model supports *H2* via the impact of financial assets. We find that a 1% increase in financial assets relative to total assets reduces intangibles assets by 3.8%. However, financial profits relative to total profits have no significant impact on intangibles assets. The GMM model also supports *H3* through the impact of dividends payments: a 1% increase in dividend payments relative to total equity reduces intangibles assets by 1.3%. The long-run estimates of the GMM model are much larger than those from the fixed effects model; here, a 1% increase in financial assets relative to total assets reduces intangible assets by 7.7%, and a 1% increase in dividend payments relative to total equity reduces intangibles assets by 2.7%.¹⁵ Financial profits and equity repurchases

¹⁵ It is expected that the long-run coefficients will be larger than the short-run coefficients because intangibles assets are quite persistent in our data, with a coefficient of 0.5. Given the formula above, the long-run

Table 4. Estimation of the effects of financialisation on intangibles assets

	log(intangibles)			log(intangibles/TA)			
	(1)	(2)	(3)	(4)			
		FE	GMM		FE	GMM	
			Short run	Long run		Short run	Long run
$\text{Log}(\text{intangibles})_{t-1}$		0.504*** (0.092)					
$\text{Log}(\text{intangibles}/\text{TA})_{t-1}$					0.492*** (0.114)		
H1—Access to finance channel							
$(\text{FL}/\text{TA})_{t-1}$	0.005* (0.003)	0.001 (0.003)	0.002 (0.006)	0.004 (0.003)	0.001 (0.003)	0.001 (0.007)	
H2—Crowding-out channel							
$(\text{FA}/\text{TA})_{t-1}$	-0.041** (0.020)	-0.038** (0.017)	-0.077** (0.030)	-0.039** (0.019)	-0.036*** (0.013)	-0.071*** (0.021)	
$(\text{FP}/\text{TP})_{t-1}$	-0.0008** (0.000)	-0.000 (0.000)	-0.000 (0.000)	-0.0008** (0.000)	-0.000 (0.000)	-0.000 (0.000)	
H3—Shareholder value orientation channel							
$(\frac{\text{Dividends}}{\text{Equity}})_{t-1}$	-0.013* (0.007)	-0.013* (0.007)	-0.027* (0.014)	-0.011* (0.006)	-0.013*** (0.005)	-0.026** (0.012)	
$(\frac{\text{Repurchase}}{\text{Equity}})_{t-1}$	-0.019*** (0.007)	-0.044 (0.029)	-0.088 (0.057)	-0.012 (0.008)	-0.042* (0.025)	-0.082 (0.053)	
Controls							
$(\frac{\text{Interest Payments}}{\text{TA}})_{t-1}$	0.005 (0.006)	-0.002 (0.005)	-0.004 (0.011)	0.003 (0.006)	-0.002 (0.006)	-0.003 (0.011)	
$\text{Log}(\text{TA})_{t-1}$	1.280*** (0.374)	0.148 (0.644)	0.298 (1.308)	0.740** (0.363)	0.153 (0.769)	0.301 (1.518)	
2011		0.120 (0.331)			0.007 (0.377)		
2012	0.067 (0.115)	0.230 (0.387)		0.122 (0.101)	0.163 (0.401)		
2013	0.253 (0.179)	0.280 (0.447)		0.367** (0.168)	0.274 (0.448)		
2014	0.169 (0.161)	0.042 (0.298)		0.339** (0.157)	0.106 (0.330)		
2015	-0.186 (0.168)	-0.267 (0.237)		0.283* (0.167)	0.060 (0.262)		
2016	0.320 (0.236)			0.461** (0.231)			
Observations	410	304		410	304		
R-squared	0.265			0.175			
Arellano–Bond test for AR(1)		-1.75*			-1.69*		
Arellano–Bond test for AR(2)		1.411			1.233		
Hansen j statistic		63.35			55.29		
Hansen p value		0.428			0.714		
Number of instruments		77			77		

Notes: Robust standard errors in parentheses.

Source: Regression analyses based on balance sheet data for Brazilian manufacturing firms.

*** $p < 0.01$,

** $p < 0.05$,

* $p < 0.1$.

4.1 Ancillary analyses: exploring similarities and differences in the effects of financialisation on intangibles and patents

Next, we explore the effects of financialisation on firm's patenting activities. As previously discussed, intangibles capture a broader base of firm's innovativeness, whereas patents have been traditionally used to measure technological innovation especially in manufacturing sectors and in advanced economy contexts. It is therefore important to explore any differences and similarities in the ways in which financialisation influences intangible assets and patents in the DEC context. To do this, we collected patent data manually from the World Intellectual Property Organisation, consisting of the number of patent applications made by each company in a given year. [Table 5](#) shows the descriptive statistics for the patent data. Firms had applied for about 2.6 patents on average, corresponding to about 1% of industry total. The highest number of patent applications for a single firm was 193, corresponding to about 63% of industry patent applications in that year.

The first three columns of [Table 6](#) show our main results from FE and short and long-run GMM models with the logarithm of intangible assets as the dependent variable, as in [Table 4](#). The last column shows results from an OLS regression, where the dependent variable is the total number of patents a firm has applied for in each year as a proportion of the total number of patents applied for by the firms in our sample that are in the same 2-digit industry sector. This measures patent intensity relative to the industry. For consistency with our main regressions, we use one lag of all independent variables.^{19,20}

For access to finance the patent results show a positive, but insignificant impact of financial liabilities on patent intensity. For the crowding out channel, the results show that financial assets have a negative, but insignificant impact on patent intensity and financial profits have no impact. For shareholder value orientation, the patents regression shows a negative, but insignificant impact of dividends and a positive, but insignificant impact of repurchases on patent intensity. We use similar control variables as in our main models; these show a negative and significant impact of interest payments, and a positive but insignificant impact of total assets on patent intensity.

Table 5. Descriptive statistics of patents

	Mean	SD	Min	Max
Total patents	2.65	14.60	0	193
Patent intensity ^a	0.99	3.18	0	63

^a Patents as a % of industry total.

Source: Authors calculations using data from the World Intellectual Property Organisation.

¹⁹ In unreported regressions, we use contemporaneous and up to two-year lags of all independent variables, here recognising that it may take longer for financial decisions to be reflected in patent applications. The results are broadly consistent with the ones reported herein, showing no significant relationship between financialisation and patent intensity. We report the model with one lag for consistency and ease of comparison with our main specification.

²⁰ Results from tobit models for patent intensity show similar results.

Table 6. *The effects of financialisation on patents: exploring differences with intangibles*

	FE	log (Intangibles)	Patent intensity	
		GMM Short run	GMM Long run	OLS
log(intangibles)t-1		0.504*** (0.092)		
H1—Access to Finance channel (FL/TA)t-1	0.005* (0.003)	0.001 (0.003)	0.002 (0.006)	0.004 (0.003)
H2—Crowding-out channel (FA/TA)t-1	-0.041** (0.020)	-0.038** (0.017)	-0.077** (0.030)	-0.003 (0.007)
(FP/TP)t-1	-0.0008** (0.000)	-0.000 (0.000)	-0.000 (0.000)	0.000 (0.000)
H3—Shareholder value orientation channel (Dividends/equity)t-1	-0.013* (0.007)	-0.013* (0.007)	-0.027* (0.014)	-0.004 (0.014)
(RepurchaseEquity)t-1	-0.019*** (0.007)	-0.044 (0.029)	-0.088 (0.057)	0.042 (0.030)
Controls				
(Interest paymentsTA)t-1	0.005 (0.006)	-0.002 (0.005)	-0.004 (0.011)	-0.010*** (0.003)
log(TA)t-1	1.280*** (0.374)	0.148 (0.644)	0.298 (1.308)	0.388 (0.341)
2011		0.120 (0.331)		
2012	0.067 (0.115)	0.230 (0.387)		0.359 (0.296)
2013	0.253 (0.179)	0.280 (0.447)		0.817** (0.346)
2014	0.169 (0.161)	0.042 (0.298)		0.252 (0.267)
2015	-0.186 (0.168)	-0.267 (0.237)		0.135 (0.344)
2016	0.320 (0.236)			0.071 (0.323)
Observations	410	304		497
R-squared	0.265			0.043
Arellano–Bond test for AR(1)		-1.75*		
Arellano–Bond test for AR(2)		1.411		
Hansen j statistic		63.35		
Hansen p value		0.428		
Number of instruments		77		

Notes: Robust standard errors in parentheses.

Source: Regression analyses based on balance sheet data for Brazilian manufacturing firms and data from the World Intellectual Property Office.

*** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

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Appendix A: Sectoral variations in intangible assets and financialisation variables

There is a good variation of the levels of intangible assets, as well as the financial variables across sectors (see [Table A1](#)). Sectors with relatively high intangible assets, on average, are miscellaneous manufacturing (13%) and food manufacturing (10%). Firms engaged in the manufacturing of leather, furniture, paper, beverages and tobacco, and transport equipment all have levels of intangible assets of 3–6% of total assets, consistent with the overall sample average (5%).

For financialisation variables, firms engaged in the manufacturing of Computer and Electronic products and Textiles have the highest level of financial liabilities, more than double their total assets; sectors with the lowest financial liabilities are Beverage and Tobacco (5% of total assets) and Electrical Equipment (3% of total assets). Sectors with highest financial assets are those engaged in the manufacturing of Leather (50% of total assets) and Machinery (47% of total assets). Sectors with the lowest level of financial assets are Paper manufacturing (16% of total assets) and Wood manufacturing (14% of total assets). The highest levels of financial profits relative to total assets are in Apparel manufacturing (163%), Wood manufacturing (87%) and Food manufacturing (85%); sectors with the greatest financial losses are Machinery manufacturing and Paper manufacturing, both making average financial losses of around five times their total assets over the sample period. Sectors that paid the most dividends, on average,

Table A1. Means of variables by two-digit SIC sector

Two digit SIC sector	Frequency	% of total	Intangibles/TA	FL/TA	FA/TA	FP/TP	Dividends/equity	Repurchase/equity	Interest payments/TA	log(TA)
Apparel manufacturing	30	7.32	0.031	0.254	0.243	1.631	-0.042	-0.002	5.945	-0.198
Beverage and tobacco product manufacturing	6	1.46	0.051	0.048	0.212	-0.134	-0.192	-0.004	10.157	-0.062
Chemical manufacturing	28	6.83	0.048	0.277	0.269	0.338	-0.032	0	6.692	-0.127
Computer and electronic product manufacturing	5	1.22	0.001	2.262	0.205	0.141	0	-0.066	4.08	-0.132
Electrical equipment, appliance, and Co fabricated metal product manufacturing	6	1.46	0.001	0.032	0.37	-1.674	-0.031	0	3.922	-0.008
Fabricated metal product manufacturing	46	11.22	0.026	0.297	0.296	0.511	-0.021	-0.002	4.818	-0.14
Food manufacturing	50	12.2	0.098	0.35	0.285	0.848	-0.026	-0.013	7.466	-0.197
Furniture and related product manufacturing	9	2.2	0.055	0.204	0.25	-1.711	-0.036	-0.001	6.851	-0.055
Leather and allied product manufacturing	27	6.59	0.058	0.266	0.51	0.218	-0.057	-0.003	6.27	-0.172
Machinery manufacturing	8	1.95	0.009	0.385	0.467	-5.854	-0.069	-0.004	7.964	-0.153
Miscellaneous manufacturing	43	10.49	0.129	0.292	0.351	0.038	-0.03	-0.006	5.508	-0.155
Non-metallic mineral product manufacturing	11	2.68	0.007	0.21	0.27	-1.917	-0.079	0	5.648	-0.174
Paper manufacturing	25	6.1	0.052	0.34	0.157	-4.603	-0.025	-0.001	7.969	-0.092

Table A1. *Continued*

Two digit SIC sector	Frequency	% of total	Intangibles/TA	FL/TA	FA/TA	FP/TP	Dividends/equity	Repurchase/equity	Interest payments/TA	log(TA)
Plastics and rubber products manufacturing	3	0.73	0.002	0.153	0.238	0.587	0	0	4.875	-0.045
Primary metal manufacturing	21	5.12	0.012	0.215	0.245	0.246	-0.018	-0.002	7.483	-0.083
Textile mills	23	5.61	0.005	1.027	0.251	0.254	-0.012	0	4.145	-0.16
Transportation equipment manufacturing	62	15.12	0.049	0.399	0.305	0.174	-0.04	0	6.583	-0.153
Wood product manufacturing	6	1.46	0.000	0.137	0.139	0.875	-0.002	0	6.595	-0.08
Total	410	100	0.052	0.361	0.295	-0.113	-0.034	-0.004	6.29	-0.145

are Beverage and Tobacco (19% of total assets), Non-metallic Mineral Products (8% of total assets), and Machinery manufacturing (7% of total assets); sectors that paid no dividends over the sample period are Plastics and Rubber manufacturing and Computer and Electronic product manufacturing. Sectors with the highest equity repurchases are Computer and Electronic Product manufacturing (6.6% of total assets) and Food manufacturing (1.3% of total assets), whereas various sectors made no repurchases over the sample period, including those engaged in the manufacturing of Non-metallic Mineral Products, Electrical Equipment, Textiles, Wood and Plastics and Rubber.

Appendix B: Robustness to normalising by total assets

	GMM									
	FE		(3)		(4)		(5)		(6)	
	Log intangibles	Log (intangibles/TA)	Fully normalised by TA	Log intangibles	Log (intangibles/TA)	Fully normalised by TA	Short run	Long run	Short run	Long run
$Log(intangibles)_{t-1}$				Short run	Long run	Short run	0.504*** (0.092)	Long run	Short run	Long run
$Log(intangibles/TA)_{t-1}$				Short run	Long run	Short run	0.492*** (0.114)	Long run	Short run	Long run
$(FL/TA)_{t-1}$	0.005* (0.003)	0.003 (0.002)	0.001 (0.002)	0.001 (0.003)	0.002 (0.006)	0.001 (0.003)	0.001 (0.003)	0.001 (0.007)	0.001 (0.003)	0.002 (0.005)
$(FA/TA)_{t-1}$	-0.041** (0.020)	-0.039* (0.021)	-0.037* (0.021)	-0.038** (0.017)	-0.077** (0.030)	-0.036*** (0.013)	-0.036*** (0.013)	-0.071*** (0.021)	-0.038** (0.017)	-0.069*** (0.022)
$(FP/TP)_{t-1}$	-0.0008** (0.000)	-0.0008** (0.000)	-0.000 (0.000)	-0.000 (0.000)	-0.000 (0.000)	-0.000 (0.000)	-0.000 (0.000)	-0.000 (0.000)	-0.000 (0.000)	-0.000 (0.000)
$(FP/TA)_{t-1}$										
$\left(\frac{Dividends}{Equity}\right)_{t-1}$	-0.013* (0.007)	-0.012* (0.006)	-0.013* (0.007)	-0.013* (0.007)	-0.027* (0.014)	-0.013*** (0.005)	-0.013*** (0.005)	-0.026** (0.012)	-0.030* (0.017)	-0.054 (0.037)
$\left(\frac{Dividends}{TA}\right)_{t-1}$										
$\left(\frac{Repurchase}{Equity}\right)_{t-1}$	-0.019*** (0.007)	-0.018** (0.008)	-0.030 (0.019)	-0.044 (0.029)	-0.088 (0.057)	-0.042* (0.025)	-0.042* (0.025)	-0.082 (0.053)	-0.030* (0.017)	-0.054 (0.037)
$\left(\frac{Repurchase}{TA}\right)_{t-1}$										
$\left(\frac{Interest\ Payments}{TA}\right)_{t-1}$	0.005 (0.006)	0.003 (0.006)	0.005 (0.006)	-0.002 (0.005)	-0.004 (0.011)	-0.002 (0.006)	-0.002 (0.006)	-0.003 (0.011)	-0.001 (0.004)	-0.001 (0.007)

FE	GMM						
	(1)	(2)	(3)	(4)	(5)	(6)	
Log intangibles	Log intangibles (TA)	Log intangibles/TA	Fully normalised by TA	Log intangibles	Log (intangibles/TA)	Fully normalised by TA	
				Short run	Long run	Short run	Long run
$Log(TA)_{t-1}$	1.280*** (0.374)	0.465* (0.265)		0.148 (0.644)	0.298 (1.308)	0.153 (0.769)	0.301 (1.518)
2011				0.120 (0.331)		0.007 (0.377)	0.064 (0.125)
2012	0.067 (0.115)	0.122 (0.101)	0.091 (0.106)	0.230 (0.387)		0.163 (0.401)	0.216** (0.102)
2013	0.253 (0.179)	0.367** (0.168)	0.306* (0.159)	0.280 (0.447)		0.274 (0.448)	0.321** (0.156)
2014	0.169 (0.161)	0.339** (0.157)	0.230 (0.166)	0.042 (0.298)		0.106 (0.330)	0.139* (0.081)
2015	-0.186 (0.168)	0.283* (0.167)	0.135 (0.175)	-0.267 (0.237)		0.060 (0.262)	0.085 (0.070)
2016	0.320 (0.236)	0.461** (0.231)	0.069 (0.193)				
Observations	410	410	410	304	304	304	304
R-squared	0.265	0.144	0.132				
Arellano-Bond test for AR(1)				-1.75*		-1.69*	-1.66*
Arellano-Bond test for AR(2)				1.411		1.233	1.180
Hansen j statistic				63.35		55.29	55.50
Hansen p value				0.428		0.714	0.418
Number of instruments				77		77	68

Notes: Robust standard errors in parentheses.
 *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.