Exploring the effect of professionalization, risk-taking and technological innovation on business performance

Francisco García-Lopera, José Manuel Santos-Jaén, Mercedes Palacios-Manzano, Daniel Ruiz-Palomo

1 Department of Applied Economics (Mathematics), University of Malaga, Malaga, Spain, 2 Department of Accounting and Finance, University of Murcia, Murcia, Spain, 3 Department of Finance and Accounting, University of Malaga, Malaga, Spain

These authors contributed equally to this work.

Abstract

The aim of this paper is to analyze the influence of professionalization over firm’s performance and the effect of two mediating variables, risk-taking and technological innovation. A total of 310 Spanish SMEs were surveyed, and the study was conducted using partial least squares path modelling (PLS-SEM) technique. The findings showed that firm’s performance is influenced by professionalization, risk-taking and technological innovation. These effects are not only direct and positive, but there are also important indirect effects that reinforce the positive effects of professionalization on firm’s performance. This research contributes to the literature on professionalization considering mediating effects of risk-taking and technological innovation in the relationship between professionalization and firm’s performance. The results provide interesting implications for theory and practice, indicating how companies can orient their strategies with the aim of gaining competitive advantage in order to increase their performance.

1. Introduction

In the current economic and social climate, companies, especially Small and Medium-sized Enterprises (SMEs), have to maximise their growth opportunities, so they have to direct their strategies towards seeking competitive advantages that will allow them to differentiate themselves from their competitors. This search for competitive advantages requires an adequate professionalization of the companies. As Diéguez-Soto et al. [1] have already argued, professionalization increases profitability and achieves competitive advantages for firms. Therefore, professionalization is recognized as a need for firms to be successful [2] and achieve improved financial and non-financial performance.

According to Kaplan and Norton [3], the professionalization of the company through the improvement of internal processes determines the critical organisational activities that will enable the company to improve its productivity and achieve differentiated value in order to
achieve its development objectives. After successful professionalization, companies achieve professional management with higher degrees of structure and defined processes [4].

Based on an appropriate professionalisation, companies must direct their strategy towards launching new market offerings, take risks as a result of trying new products, services, or markets, and be more proactive than rivals [5]. At this point, risk-taking appears as the firm characteristic to take risks by strategically considering the business-related opportunities in uncertain situations. According to Soto-Acosta et al. [6], Technological innovation is the strategy defined for the transformation of ideas and knowledge into these new products and processes.

Previous literature is very prolific in studying the relationship of these variables to each other. However, it is not common to find studies as a whole. Thus, many studies analyse the influence of professionalization on performance [7–9] or risk-taking on innovation [10, 11]. However, the goal of this paper is to analyze how both risk-taking and technological innovation mediate the effect of professionalization on SMEs’ performance. The key research questions we are trying to answer are: Does professionalization influence performance in SMEs?; Is this relationship mediated by risk-taking or technological innovation?; Is the effect of professionalization on performance mediated by risk-taking?. With this purpose, a Structural Equations Modeling based on Partial Least Squares (PLS-SEM) has been developed in a sample of 310 Spanish SMEs. PLS-SEM is a suitable technique for solving problems even when very complex relationships exist because the optimisation algorithm maximises the explained variance of the model’s independent variables [12]. In this way, PLS-SEM has allowed us to understand the causal relationships between latent variables [13].

The findings in this paper have demonstrated the effect of professionalization on firm performance and the mediating effect of risk-taking and technological innovation on this relationship. For this reason, this paper contributes to previous research by demonstrating how SMEs can only expect to improve their performance if they first carry out an adequate professionalization process that allows them to implement a risk-oriented strategy and to implement appropriate technological innovation.

This paper continues in section 2, showing the hypotheses development. In section 3, the research methodology is established. The results obtained after applying PLS-SEM are presented in section 4. Finally, section 5 concludes with a discussion of these results and a description of the practical and theoretical implications and limitations.

2. Literature review

The reason for human resource professionalization is established in agency theory, which theorizes that managers will follow self-interested goals, rather than the owner’s goals, if their behavior is not monitored [14]. From an agency perspective, combined ownership and management may reduce the threat of agency problems related to information asymmetries and managerial appropriations [15].

Several studies have examined the impact of professionalization on firm performance, and in general, the results show that professionalization improves financial performance. [7–9]. Most previous research on the professionalization of business have been carried out in family firms [7, 9, 16] and Western economies [17].

Madison et al. [8] conducted research to investigate the treatment of employees in family and non-family organizations and how it affects firms’ performance. Results revealed that human resource professionalization is positively associated with the performance of the family firm. In this line, Lien and Li [16] showed that, adopting non-family management in post-IPO family firms harms firm performance due to weak governance institutions in Taiwan.
However, more research on non-family employees is needed. For this reason, we establish the following hypothesis:

\( H_1 \): Professionalization is positively related to performance.

The professionalization process encompasses many different aspects that a firm must address [9]; among them is the definition of the level of Risk-taking that the company is willing to undertake.

Professionalization enables the company to better understand and assess the developments around it. This will enable it to react much more quickly to changes in the environment. Therefore, the company is willing to accept larger venturing risks [18] since the company will feel more able to take risks [19]. In the same vein, a low level of professionalization in business makes it difficult to step out of one’s comfort zone and adopt Risk-taking [20].

Previous studies have shown that the more professionalized a company is, the more Risk-taking it is likely to be [20–23]. In this respect, it has been shown that a long-term orientation, through better use of resources and skills following successful professionalization, can provide a favourable context for increased entrepreneurial Risk-taking [24].

Literature has recognised Risk-taking as the central feature of entrepreneurship and a contributor to performance [25]. Based on one of the most fundamental asset pricing theories in financial economics theory, the Capital Asset Pricing Model [26], Risk-taking directly influences performance, through the risk-return trade-off since entrepreneurs will only take on the riskier or more uncertain ventures/strategies if they are accompanied by a higher expected return [25]. Moreover, Risk-taking provides companies the capabilities to transform themselves in response to changes in the environment, with the aim of obtaining a competitive advantage and ensuring long-term survival [27]. This is because a risk orientation allows companies to introduce new products and brands ahead of their competitors [28]. In the same vein, Hoskisson et al. [29], claimed that Risk-taking is vital for managers to compete in a dynamic market and to respond the competitive threats. Similarly, Putniņš and Sauka [25] demonstrated that constructive Risk-taking is the central driver of company performance, mirroring the principle of risk and return in financial investment settings. Hence, it can be argued that as firms invest in new projects, they take financial risk and greater risk conditions will result in higher financial performance [30].

Based on the above, previous studies have shown the positive effect of Risk-taking on performance [30–32].

Professionalization helps firms cope with their competitive environment, increase strategic decision-making quality, and thus increase the firm’s performance [33–35]. At the same time, risk-taking directly correlates with performance, which can be understood through the risk-return trade-off central to financial economics theory [25].

We further propose that professionalization positively contributes to performance by impacting the types of risk taken within the firm. For this reason, it is to be expected that Risk-taking mediates the relationship between professionalization and performance (a mediated relationship).

Given the above, we establish the following research hypothesis:

\( H_2 \): Risk-taking partially mediates the relationship between professionalization and performance.

This \( H_2 \) hypothesis is sub-divided into the following three hypotheses:

\( H_{2a} \): Professionalization has a positive effect on Risk-taking.

\( H_{2b} \): Risk-taking has a positive effect on performance.
Professionalization indirectly affects performance through Risk-taking.

Professionalization supports technological innovation across its initiation, adoption and implementation [1]. Therefore, technological innovation in the firm may be conditioned by the knowledge that the firm can contribute during the innovation process [36]. In this sense, Liang et al., [37] stated that the performance of technological innovation depends mostly on the process of knowledge creation. Thus, the greater the professionalization of the firm, the greater its capacity to innovate is.

According to [38], the professionalization of a company gives it advantages in terms of new product development. For example, the more professional the company’s CEO is, the greater the capacity to identify opportunities for change and of developing strategic planning and management through the innovation of products, services, or processes [39].

Moreover, according to Resources-Based View (RBV), professionalization provides firms with resources, ideas, labels and visions with which to build and develop the firm, thereby increasing the incentive to invest in innovation [40], as well as affecting the innovation capacity of the whole organisation [41]. Professionalization improves internal processes reducing cost or increasing quality and reliability [1].

Based on the above, it is expected that firm professionalization will develop better strategic planning and management through the innovation of products and processes.

The concept of innovation can be defined as new knowledge and ideas transformed into new products and/or services, new technologies, new processes and new organizational structures [42].

Technological innovation is a key factor in business performance and plays an important role in the corporation’s growth. SMEs have to monitor their competitive position through innovation [43]. It is necessary to raise competitiveness capability and intensify firm efficiency and productivity [44].

The topic of the impact of innovation on firm performance has been studied by previous researchers [11, 45–47]. Relevant studies reveal that the two concepts are positively and significantly correlated [48–51].

A firm’s knowledge capabilities, not only effectively but also innovatively, allow it to improve its performance [6, 52]. According to Darroch [52], knowledge management can be considered a synchronizing tool used to translate these assets into capabilities that enhance organizational performance. Similarly, Soto-Acosta et al. [42] argued that technological innovation provides firms with a strategic orientation to achieve sustainable competitive advantage. This explains why technological innovation has become an essential factor that contributes to business performance.

This is because companies that develop more innovative products and services achieve benefits over their competitors [50] because innovative products and services face less competition to be introduced in the market, allowing the company to increase profits and differentiate itself from the competition [53–55].

As has been established above, professionalization provides companies with greater capabilities to carry out appropriate technological innovation. This technological innovation, in turn, enables firms to gain competitive advantages that allow them to differentiate themselves from their competitors and increase their performance. It is, therefore, to be expected that an increase in the professionalization of firms will, in turn, have an indirect and positive effect on performance by increasing innovative capacity.

Given the above, we establish the following research hypothesis:

\[ H_3c: \text{ Technological innovation partially mediates the relationship between professionalization and performance.} \]
This H₃ hypothesis is sub-divided into the following three hypotheses:

H₃a: *Professionalization has a positive effect on Technological innovation.*

H₃b: *Risk-taking has a positive effect on performance.*

H₃c: *Professionalization indirectly affects performance through Technological innovation.*

The link between Risk-taking and innovation has been proven by several scholars [10, 11]. Technological innovation involves not only an initial expenditure on R&D but also uncertain benefits [25]. For its part, Risk-taking indicates the companies’ willingness to invest resources in technological innovation [56]. In the same vein, Miller and Friesen [5] suggest that Risk-taking is “entrepreneurial” when it is associated with innovation.

An adventurous entrepreneurial spirit in business without fear of risk and in search of higher returns facilitates experiments, among which are acquiring, learning and absorbing new external technology [57]. In accordance with the preceding, Mao and Zhang [58] state that Risk-taking is an important driver of firm technological innovation.

The mediating effect of innovation on the relationship between Risk-taking and performance has been studied in the literature [59]. Regarding technological innovation, the resource-based view of the firm suggests that technological innovation mediates the relationship between Risk-taking and performance [60]. Based on this theory, when resources are constrained, companies might be forced to combine the competing demands and engage in innovative Risk-taking, directing this Risk-taking towards innovative activities [25]. Therefore, this indirect approach consists of improving companies’ innovation capabilities. In this line, Jeon [61] proved that technological innovation has a mediating effect on the relationship between Risk-taking and performance.

In the same way, on the basis of the established above, the professionalization of firms can be expected to have a spill-over effect on technological innovation. This professionalization will not only provide them with the necessary knowledge with which to carry out the implementation of technological innovation successfully. However, it will also allow them to have the capacity to assume the risks that any technological innovation process entails. For this reason, it is to be expected that Risk-taking mediates the relationship between professionalization and technological innovation.

Based on the information provided above, we state the following hypotheses:

H₄: *The effect of Risk-taking on performance is partially mediated by Technological innovation.*

H₅: *Risk-taking partially mediates the relationship between Professionalization and Technological innovation.*

The H₄ hypothesis is sub-divided into the following two hypotheses:

H₄a: *Risk-taking is positively associated with Technological innovation.*

H₄b: *Risk-taking indirectly affects performance through Technological innovation.*

On the basis of all the above, it seems reasonable to believe that by increasing professionalization, Risk-taking increases the capacity of companies to carry out innovative activities, which has an impact on their performance. Therefore, a sequential mediation of the relationship between professionalization and performance by Risk-taking and Technological innovation can be expected. Hence, we propose the following hypothesis:
Risk-taking and Technological innovation sequentially mediate the relationship between Professionalization and performance.

In order to test the hypotheses put forward, in Fig 1 our proposed model is depicted. It shows four constructs, an exogenous construct (professionalization) and three endogenous constructs (Risk-taking, technological innovation and performance).

2.1. Sample

This study was carried out with a sample of 310 Spanish SMEs. For this purpose, the Central Business Directory of the National Statistical Institute (INE) was used. The sample was segmented by activity and size. The distribution of the sample is presented in Table 1. The sample was selected in SABI database through the principles of stratified sampling for finite populations. The estimation of the sample considered a relative frequency of answers in a specific item is $p = 0.5$, to a maximum error of 5.6% at a confidence level of 95%. To confirm that statistically significant relationships will be identified in the proposed model and that the sample size is sufficient to carry out the research, a post hoc analysis has been conducted using the G*Power Version 3.1.9.4 software tool [62]. Assuming a standard error of 0.05 and an effect size of 0.15, we have calculated the statistical power of the sample, which is 1 (over than the shortcut value of 0.8 [63]). Therefore the sample size is acceptable.

The information was obtained through telephone surveys of company managers from October 2016 to January 2017. The survey was carried out personally with managers, as they are the ones who know best the objectives, strategies, concerns of the companies [64]. According to Fisher [65], to reduce social acceptance bias, the survey was anonymously undertaken. The nonresponse bias has been analysed through the t-test and the chi-squared test for all the

```
Table 1. Sample composition.

| Size (Number of employees) | Total of companies | Micro companies (<10) | Small companies (10–49) | Medium companies (50–250) |
|---------------------------|--------------------|-----------------------|-------------------------|---------------------------|
| Number                    | Percent of total   | Number                | Percent of total         | Number                    | Percent of total |
| 310                       | 100                | 111                   | 35.80%                  | 130                       | 41.90%           | 69               | 11.80%         |

| Activity                  | Total of companies | Manufacturing | Construction | Trade & Commerce | Services |
|---------------------------|--------------------|--------------|--------------|------------------|----------|
| Number                    | Percent of total   | Number       | Percent of total | Number           | Percent of total | Number       | Percent of total |
| 310                       | 100                | 98           | 31.60%       | 90               | 29.00%       | 60           | 19.40%        | 62               | 20.00%          |

Source: Authors

https://doi.org/10.1371/journal.pone.0263694.t001
```
constructs. For this purpose, the responses have been divided into two groups, a first group containing the first responses (75% of the responses) and a second group containing the last responses (25% of the responses). The results show that there are no problems concerning the nonresponse bias. Finally, a standard method bias has been discarded when analysing the results of the variance inflation factors (VIF). The results, which will be shown in the following section, show how the VIF varies from 1.525 to 2.608. Thus, all the values are smaller than 3.3, common method bias is not a problem in this research [66].

2.2. Measures

Based on previous studies, the four variables that make up the proposed model have been developed. All the indicators that make up the latent variables are measured through 5-points Likert scale. Table 2 shows the definition and composition of the variables.

Table 2. Variables used in the research.

| Professionalization | In relation to the professionalization of internal processes in the company, please indicate your degree of agreement with the following statements |
|---------------------|------------------------------------------------------------------------------------------------------------------------------------|
| Prof_1              | There is a formally established organisational structure                                                                          |
| Prof_2              | There is a formalised staff performance and incentive system in place                                                             |
| Prof_3              | There is an annual schedule and follow-up of management team meetings                                                           |

| Risk-taking         | In relation to the risk-taking propensity, please indicate your level of agreement with the following statements |
|---------------------|------------------------------------------------------------------------------------------------------------------|
| Risk_1              | I have a strong propensity for high-risk projects                                                                    |
| Risk_2              | I believe that knowing the environment, bold and far-reaching actions are necessary to achieve the company’s objectives |
| Risk_3              | When faced with decision-making under conditions of uncertainty, I normally adopt a bold and aggressive stance in order to maximise the probability of exploiting potential opportunities. |

| Technological innovation | The evolution of your company over the last two years, and compared to other companies in your sector, can be rated in relation to your company’s products and services |
|--------------------------|-----------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| Tech_1                   | The number of new products or services introduced by your company per year                                                                                                                                |
| Tech_2                   | The pioneering nature of your company’s introduction of new products or services                                                                                                                        |
| Tech_3                   | The speed of response to the introduction of new products or services by other companies in the sector, can be qualified in relation to the following processes |

| Performance              | Please indicate your level of agreement with the following statements, taking the competition as a reference |
|--------------------------|------------------------------------------------------------------------------------------------------------------|
| Perf_1                   | Offers higher quality products                                                                                   |
| Perf_2                   | Has more efficient internal processes                                                                             |
| Perf_3                   | Has more satisfied customers                                                                                     |
| Perf_4                   | Adapts earlier to market changes                                                                                |
| Perf_5                   | It is growing more                                                                                            |
| Perf_6                   | It is more profitable                                                                                           |

Source: Authors

https://doi.org/10.1371/journal.pone.0263694.t002
2.2.1. Professionalization. In order to be able to measure the degree of professionalization of the companies, respondents were asked to answer from 1 (strongly disagree) to 5 (strongly agree) on the existence of an organisational structure, the existence of a system of incentives and employee performance and the frequency and scheduling of management meetings [8, 17, 67].

2.2.2. Risk-taking. The risk-taking dimension has been created through the three items developed by Miller and Friesen [5]. For this purpose, respondents were asked to answer from 1 (strongly disagree) to 5 (strongly agree) about the propensity for high-risk projects, the relationship between the execution of courageous actions and business objectives, and the adoption of a courageous stance in situations of uncertainty in order to exploit potential opportunities.

2.2.3. Technological innovation. According to Damanpour and Gopalakrishnan [68], two types of technological innovation have been identified: process innovation and product innovation. Consistent with previous studies [69, 70], six indicators have been used to create this variable. For this purpose, respondents were asked to answer from 1 (strongly disagree) to 5 (strongly agree) about the introduction of new products and services and on the implementation of new processes by the company.

2.2.4. Performance. In order to measure the financial and non-financial performance of companies, a latent variable has been created, consisting of six indicators [6, 71]. For this purpose, respondents were asked to answer, concerning the competence, from 1 (strongly disagree) to 5 (strongly agree) about the quality of products, efficiency of internal processes, customer satisfaction, its ability to adapt to change and its growth and profitability.

3. Results

3.1. Data analysis

Employing Smart-PLS software 3.3 [72], this research has applied partial least squares equation modelling (PLS-SEM) [73]. As this model presents a mix of formative (professionalization) and reflective factors (risk-taking, technological innovation and performance), PLS consistent has been used [74]. In order to obtain standard errors and t-statistics to evaluate the model, a bootstrap method of resampling of 10,000 has been used [75].

We have chosen PLS-SEM to run our model because the model is formed by four composite type A [76]. In addition, PLS-SEM is an ideal technique when the relationships analysed are very complex, especially if there are mediating effects [77, 78], it does not require a very large sample size and the data does not have to be strictly standardised [79].

3.2. Overall model: Test of goodness-of-fit (GoF)

As our model has a confirmatory purpose, we have started the analysis of the estimated model by focusing on several rates of overall goodness of fit (GoF) established by Henseler [13]. The results are shown in Table 3.

The outcome for the standardized root mean square residual (SRMR) is 0.042, which is considerably below the maximum limit of 0.08 established [13]. Moreover, various model fit analyses (SRMR, dULS, dG) have been carried out using bootstrap-based inference statistics. In Table 3 it can be seen how all the results are under the bootstrap-based 99% (HI99) percentile. Therefore, the discrepancy observed between the empirical correlation matrix and the one implied by the model is not significant. Hence, given the results obtained, it can be stated an excellent model fit [80].
3.3. Measurement model

This model is made up of one reflective construct (professionalization) and three formative constructs (risk-taking, technological innovation and performance). According to Hair et al. [66], the evaluation of the formative dimensions is not the same as for the reflective-type dimensions. The evaluation of the reliability and validity is not applicable in formative construct because they do not need to be correlated [81, 82].

For the reflective construct, in order to validate the measurement model, traditional measures of internal consistency, reliability and validity has been verified. For this reason, the factor loadings, Cronbach’s Alpha, composite reliability [83], the Dijkstra-Henseler rho ratio [84] and the average variance extracted (AVE) has been analysed.

According to Valls Martínez et al. [85], these measures are determined as follows:

**Composite reliability**, which should range from 0.7 to 0.95, is the lower limit of internal consistency reliability of the reflective construct. This measure is determined by:

\[
\rho_c = \frac{\left( \sum_{k=1}^{K} l_k \right)^2}{\left( \sum_{k=1}^{K} l_k \right)^2 + \sum_{k=1}^{K} \text{var}(e_k)}
\]

where \( l_k \) is the outer loading of the manifest variable \( k \) corresponding to a latent variable measured with \( K \) indicators; \( e_k \) is the measurement error of \( k \); and \( \text{var}(e_k) \) corresponds to the measurement error variance and it is calculated as \( 1 - l_k^2 \).

Cronbach’s alpha is the upper limit of internal consistency reliability:

\[
\text{Cronbach’s } \alpha = \frac{K \bar{\tau}}{1 + (K - 1) \bar{\tau}},
\]

where \( \bar{\tau} \) is the mean of the triangular correlation matrix.

The Dijkstra-Henseler’s Rho usually stands between the two previous measures [73]:

\[
\rho_A = \frac{(\hat{w}'\hat{w})^2 \hat{w}'(S - \text{diag}(S))\hat{w}}{\hat{w}'(\hat{w}\hat{w}' - \text{diag}(\hat{w}\hat{w}'))\hat{w}},
\]

where \( \hat{w} \) is the estimated weight vector of the construct, and \( S \) is the empirical covariance matrix of the manifest variables.

---

**Table 3. Test of model fit.**

|                     | Estimated Model | Saturated Model |
|---------------------|----------------|-----------------|
| Value               | HI99           | Value           | HI99           |
| SRMR                | 0.042          | 0.044           | 0.042          | 0.044 |
| \( d_{ULS} \)       | 0.302          | 0.332           | 0.302          | 0.329 |
| \( d_c \)           | 0.101          | 0.126           | 0.101          | 0.126 |

Standardized root mean square residual (SRMR). Unweighted least squares discrepancy (\( d_{ULS} \)). Geodesic discrepancy (\( d_c \)).

https://doi.org/10.1371/journal.pone.0263694.t003
The average variance extracted (AVE) is a measure of the convergent validity [86]:

$$\text{AVE} = \frac{\sum_{k=1}^{K} \bar{\lambda}_k^2}{K}.$$  

It is considered acceptable when its value exceeds 0.5, which means that the construct explains more than 50% of its manifest variables variance.

The results of the tests carried out are shown in Table 4. All results exceed the minimum values established [87], except for the loading of one indicator, although its value close to 0.7 is acceptable, supporting the reliability and convergent validity for the construct and its dimensions.

For the formative constructs the significance and relevance of items as well as the Variance Inflation Factor (VIF) to exclude problems of collinearity, have been evaluated. This measure

| Table 4. Measurement model results. |
|-------------------------------------|
| Mean | Loading | t-student* | $Q^2$ | $\alpha$ | $\rho_A$ | $\rho_C$ | AVE |
| Professionalization | | | | | | | |
| Prof_1 | 4.019 | 0.673 | 10.7130 | 0.761 | 0.764 | 0.762 | 0.516 |
| Prof_2 | 2.994 | 0.757 | 12.9340 | 0.764 | 0.761 | 0.762 | 0.516 |
| Prof_3 | 3.516 | 0.723 | 11.6900 | 0.762 | 0.761 | 0.762 | 0.516 |
| Mean | Weights | t-student* | $Q^2$ | VIF |
| Risk-taking | | | | |
| Risk_1 | 2.077 | 0.411 | 10.020 | 0.063 | 1.525 |
| Risk_2 | 3.203 | 0.422 | 11.908 | 0.047 | 1.679 |
| Risk_3 | 3.023 | 0.367 | 9.732 | 0.058 | 1.669 |
| Technological innovation | | | | |
| Tech_1 | 4.061 | 0.179 | 7.896 | 0.098 | 1.726 |
| Tech_2 | 3.526 | 0.265 | 9.692 | 0.206 | 1.660 |
| Tech_3 | 4.039 | 0.126 | 4.412 | 0.035 | 1.648 |
| Tech_4 | 3.742 | 0.237 | 10.512 | 0.171 | 1.833 |
| Tech_5 | 3.381 | 0.249 | 12.846 | 0.190 | 2.463 |
| Tech_6 | 3.152 | 0.254 | 11.240 | 0.198 | 2.396 |
| Performance | | | | |
| PERF_1 | 3.200 | 0.168 | 8.953 | 0.091 | 1.592 |
| PERF_2 | 3.326 | 0.226 | 13.936 | 0.192 | 1.962 |
| PERF_3 | 3.094 | 0.211 | 13.361 | 0.133 | 2.170 |
| PERF_4 | 3.119 | 0.211 | 13.075 | 0.153 | 1.965 |
| PERF_5 | 3.129 | 0.251 | 16.299 | 0.233 | 2.608 |
| PERF_6 | 2.965 | 0.216 | 14.536 | 0.154 | 2.586 |

Significance and standard deviations (SD) performed by 10,000 repetitions Bootstrapping procedure. $Q^2_B$: cross-validated redundancies index performed by a 9-step distance-blindfolding procedure. $\alpha$: Chronbach’s alpha; $\rho_A$: Dijkstra–Henseler’s composite reliability; $\rho_C$: Jöreskog’s composite reliability; AVE: Average Variance Extracted; VIF: Variance Inflation Factor

*: All the loadings and weights are significant at a 0.001 level.

Source: Authors

https://doi.org/10.1371/journal.pone.0263694.t004
is determined by [85]:

$$VIF_k = \frac{1}{1 - R^2_k}.$$ 

The results in Table 4, show how all the weights are significant and VIF values are below 3, verifying the absence of collinearity issues [87].

Based on the results obtained for both types of constructs it can be stated that the model is well formed.

In addition, the predictive relevance of the model has been confirmed through a blindfolding procedure (omission distance of 9), where all the $Q^2$ values are above 0 [88].

Finally, the satisfactory explanatory qualities of the model have been tested. For this purpose the predictive relevance of the exogenous construct has been evaluated. So, the $Q^B^2$ statistical test (a cross-validated redundancy index), has been carried out by the blindfolding method [89]. The results in Table 4 reveal that all $Q^B^2$ are greater than zero, confirming this quality [90].

### 3.4. Structural model

Before testing the hypotheses, the existence of any multicollinearity problem has been ruled out by analysing the variance inflation factor (VIF). As can be seen in Table 5, the VIF values are between 1 and 1.468. This means that these values are well below the recommended maximum limit of 3.3 [91] and therefore there are no multicollinearity problems in this model.

According to Hair et al. [92], the structural model has been analysed through the study of the magnitude, significance, algebraic sign and the effect size index ($f^2$) values of the standardized regression coefficients (path coefficients). Similarly, for the endogenous construct the determination coefficient ($R^2$) values has been analysed. For this purpose, a bootstrap re-sampling for 10,000 subsamples has been carried out.

The results in Table 5 and Fig 2 show that all hypothesis can be accepted. It can be observed that professionalization has a significant positive effect on performance, risk-taking and technological innovation, with a path coefficient value of $\beta = 0.269$, $\beta = 0.329$ and $\beta = 0.451$ respectively. Hence, these results verify the hypotheses $H_1$, $H_{2a}$, and $H_{3a}$. Risk-taking has a significant positive effect on technological innovation and performance ($\beta = 0.223$ and $\beta = 0.178$ respectively), which verifies $H_{2b}$ and $H_{4a}$. Finally, technological innovation has a significant positive effect on performance ($\beta = 0.251$), verifying $H_{3b}$.

The $R^2$ shows through the variables predicting an endogenous construct, how these can explain its variance. Therefore $R^2$ is a measure of the predictive/explanatory power of the model [93], Falk and Miller [94] established a minimum value of 0.10. The results obtained in this research show that the model explains 10.5% of the variance in risk-taking, 31.5% in technological innovation and 29.6% in performance, which demonstrates a moderate level of explanatory power for risk-taking and technological innovation.

The $f^2$ analyse the contribution of the exogenous variables to the $R^2$ of the endogenous variables. Values of 0.02, 0.15, and 0.35 indicate small, medium, and large effects [95]. As can be seen in Table 5, all links are above the minimum value of 0.02 [96], which shows its direct effect on the global performance variables. It is interesting to highlight the significant influence of professionalization on technological innovation ($f^2 = 0.266$).

### 3.5. Mediation analysis

Once the direct effects have been analysed, following the procedure laid down by [79], indirect effects have been investigated. With the aim to check the indirect effects a bootstrapping
Table 5. Results of the hypothesis testing.

| Structural paths | Path | $t$   | $f^2$ | 95CI          | $H$  | Supported |
|------------------|------|-------|-------|---------------|------|-----------|
| **Direct effects** |      |       |       |               |      |           |
| Professionalization $\rightarrow$ Performance | 0.269 | 3.365*** | 0.073 | [0.146; 0.408] | 1.419 | $H_1$ Yes |
| Professionalization $\rightarrow$ Risk-taking | 0.329 | 4.960*** | 0.121 | [0.220; 0.437] | 1.000 | $H_{2a}$ Yes |
| Risk-taking $\rightarrow$ Performance | 0.178 | 3.300*** | 0.038 | [0.088; 0.267] | 1.194 | $H_{3b}$ Yes |
| Professionalization $\rightarrow$ Technological innovation | 0.451 | 6.785*** | 0.266 | [0.344; 0.565] | 1.121 | $H_{4a}$ Yes |
| Technological innovation $\rightarrow$ Performance | 0.251 | 3.843**  | 0.062 | [0.139; 0.355] | 1.468 | $H_{5a}$ Yes |
| Risk-taking $\rightarrow$ Technological innovation | 0.223 | 3.569*** | 0.065 | [0.118; 0.322] | 1.121 | $H_{6a}$ Yes |
| **Indirect effects** |      |       |       |               |      |           |
| Individual indirect effects |      |       |       |               |      |           |
| Professionalization $\rightarrow$ Risk-taking $\rightarrow$ Performance | 0.059 | 2.792**  |       | [0.027; 0.096] | 12.854 | $H_{2c}$ Yes |
| Professionalization $\rightarrow$ Technological innovation $\rightarrow$ Performance | 0.113 | 3.698*** |       | [0.065; 0.166] | 24.619 | $H_{3c}$ Yes |
| Risk-taking $\rightarrow$ Technological innovation $\rightarrow$ Performance | 0.056 | 2.343*   |       | [0.021; 0.099] | 31.461 | $H_{4b}$ Yes |
| Professionalization $\rightarrow$ Risk-taking $\rightarrow$ Technological innovation | 0.073 | 3.312*** |       | [0.039; 0.110] | 13.931 | $H_{5b}$ Yes |
| Professionalization $\rightarrow$ Risk-taking $\rightarrow$ Technological innovation $\rightarrow$ Performance | 0.018 | 2.308*   |       | [0.007; 0.033] | 3.922  | $H_{6b}$ Yes |
| Global indirect effects |      |       |       |               |      |           |
| Professionalization $\rightarrow$ Technological innovation | 0.073 | 3.312*** |       | [0.039; 0.110] | 13.931 |          |
| Professionalization $\rightarrow$ Performance | 0.190 | 4.705*** |       | [0.125; 0.258] | 41.394 |          |
| Risk-taking $\rightarrow$ Performance | 0.056 | 2.343*** |       | [0.021; 0.099] | 31.461 |          |
| Total effect |      |       |       |               |      |           |
| Professionalization $\rightarrow$ Technological innovation | 0.524 | 9.051*** |       |               |      |           |
| Professionalization $\rightarrow$ Performance | 0.459 | 6.977*** |       |               |      |           |

$R^2$ adjusted [99% CI in brackets]: Risk-taking: 0.105 [0.450; 0.188]; Technological innovation: 0.315 [0.240; 0.414]; Performance: 0.296 [0.216; 0.413]. Blindfolding $Q^2$ index as shown in Table 4; Standardized path values reported; $f^2$: size effect index; 95CI: 95% Bias Corrected Confidence Interval; VIF: Inner model Variance Inflation Factors; VAF: Variance Accounted Formula x 100 represents the proportion mediated. Significance, t-Student, and 95% bias-corrected CIs were performed by 10,000 repetitions Bootstrapping procedure;

*: $p < 0.05$;
**: $p < 0.01$;
***: $p < 0.001$.

Only total effects that differ from direct effects are shown.

Source: Authors

https://doi.org/10.1371/journal.pone.0263694.t005

Fig 2. Results of SEM analysis.

https://doi.org/10.1371/journal.pone.0263694.g002
procedure with 10,000 samples has been utilized [66]. This method generates for the individual indirect effect and the sequential mediation 95% bias-corrected. Additionally, the size of the indirect effect in relation to the total effect has been analyzed through the variance accounted for (VAF) [73]. The results are presented in Table 5.

The findings show that risk-taking mediates the impact of professionalization on technological innovation ($\beta = 0.073^{*\ast\ast\ast}$) and performance ($\beta = 0.059^{*\ast\ast\ast}$), supporting $H_5$ and $H_{2c}$. Similarly, it can be observed that technological innovation mediates the relationship between risk-taking and performance ($\beta = 0.056^{*\ast}$), supporting $H_{4b}$, and between professionalization and performance ($\beta = 0.113^{*\ast\ast\ast}$), supporting $H_{3c}$. Moreover, a sequential indirect effect of professionalization on performance through risk-taking and technological innovation has been verified ($\beta = 0.018^{\dagger}$), supporting $H_6$.

According to Hair et al. [97], the variance accounted for (VAF) indicates the size of each of the indirect effects relative to the total effect. Therefore, the indirect effect of professionalization on technological innovation is about 13.9% of the total effect through risk-taking. Similarly, the indirect effect of professionalization on performance is about 41.39% of the total effect, with 12.85 through risk-taking, 24.6% through technological innovation, and an additional 3.92% sequentially. Finally, the indirect effect of risk-taking on performance is about 31.46% of the total effect through technological innovation. Since all effects are positive and the VAF values are below 0.8, it can be established that all mediations are partial and complementary [98].

4. Discussion and conclusion

This research aimed to analyse the effect of professionalization on firm performance, also examining the mediating effect of risk orientation and technological innovation. For this purpose, a sample of 310 Spanish SMEs has been used. The analysis was carried out using PLS-SEM.

The findings show, in line with previous research [7–9], that managers’ pursuit of their own interests means that the higher the level of professionalization of the company, the higher the performance obtained.

The results also reveal how professionalization enhances firms’ skills, allowing them to increase their Risk-taking capacity and thus better adapt to changes in the environment, which will increase their survivability and thus improve their performance. These results are in line with previous research.

On the other hand, and in line with previous research [1, 36], it has been shown that professionalization is key to being able to deal with appropriate technological innovation [43], as the latter requires the knowledge of the company. As technological innovation is a key factor in business performance, indirect professionalization affects performance by improving the innovative capacity of companies. Based on the above, it is also worth noting how it has been shown that professionalisation, by increasing the ability of companies to leave their comfort zone, indirectly influences technological innovation.

Furthermore, it has also been shown, in line with [56], that the greater capacity of companies to take risks will lead them to make greater investments aimed at increasing their technological innovation. Thus, risk orientation indirectly influences firm performance.

Finally, it is interesting to note how the results allow us to conclude the existence of a sequential influence of professionalization on risk orientation, the latter on technological innovation, resulting in a positive effect on performance.

The literature has repeatedly demonstrated the positive effect of professionalization on business performance [7–9]. Now this research shows how risk orientation and technological
innovation increase the positive effect of professionalization on performance. The results show how companies that are more committed to professionalization obtain higher performance and that their impact will be greater as they improve their Risk-taking capacity and thus technological innovation.

From a theoretical point of view, this research contributes to the business management literature by integrating the role that risk orientation and technological innovation play in the relationship between professionalization and performance. This is of vital importance to understand what strategies SMEs can develop to survive and grow in today’s changing environment.

This research has important implications for SME owners and shareholders. It has been demonstrated that a commitment to professionalization will increase company performance through the ability to take risks and carry out technological innovations to develop new products and services. Likewise, our results have shown that in order to innovate, companies have to leave their comfort zone and take risks, which undoubtedly requires prior skills that cannot be obtained without a process of business professionalisation.

This research is not without limitations, which may serve as a basis for future lines of research. This study was carried out only with Spanish SMEs, so that the results may not be extrapolated to other geographical areas. Future research could cover a more ambitious sample of SMEs from several countries. Also, this research only uses cross-sectional data, so that the results may change over time. It would be interesting for future research to use longitudinal data to analyse the effects over time.

Author Contributions

Conceptualization: Francisca García-Lopera, José Manuel Santos-Jaén, Mercedes Palacios-Manzano, Daniel Ruiz-Palomo.

Data curation: Francisca García-Lopera, José Manuel Santos-Jaén, Mercedes Palacios-Manzano, Daniel Ruiz-Palomo.

Formal analysis: Francisca García-Lopera, José Manuel Santos-Jaén, Mercedes Palacios-Manzano, Daniel Ruiz-Palomo.

Funding acquisition: Francisca García-Lopera, José Manuel Santos-Jaén, Mercedes Palacios-Manzano, Daniel Ruiz-Palomo.

Investigation: Francisca García-Lopera, José Manuel Santos-Jaén, Mercedes Palacios-Manzano, Daniel Ruiz-Palomo.

Methodology: Francisca García-Lopera, José Manuel Santos-Jaén, Mercedes Palacios-Manzano, Daniel Ruiz-Palomo.

Project administration: Francisca García-Lopera, José Manuel Santos-Jaén, Mercedes Palacios-Manzano, Daniel Ruiz-Palomo.

Resources: Francisca García-Lopera, José Manuel Santos-Jaén, Mercedes Palacios-Manzano, Daniel Ruiz-Palomo.

Software: Francisca García-Lopera, José Manuel Santos-Jaén, Mercedes Palacios-Manzano, Daniel Ruiz-Palomo.

Supervision: Francisca García-Lopera, José Manuel Santos-Jaén, Mercedes Palacios-Manzano, Daniel Ruiz-Palomo.
Validation: Francisca García-Lopera, José Manuel Santos-Jaén, Mercedes Palacios-Manzano, Daniel Ruiz-Palomo.

Visualization: Francisca García-Lopera, José Manuel Santos-Jaén, Mercedes Palacios-Manzano, Daniel Ruiz-Palomo.

Writing – original draft: Francisca García-Lopera, José Manuel Santos-Jaén, Mercedes Palacios-Manzano, Daniel Ruiz-Palomo.

Writing – review & editing: Francisca García-Lopera, José Manuel Santos-Jaén, Daniel Ruiz-Palomo.

References
1. Diéguez-Soto J, Duréndez A, García-Pérez-de-Lema D, Ruiz-Palom o D. Technological, management, and persistent innovation in small and medium family firms: The influence of professionalism. Can J Adm Sci Can des Sci l'Administ ration. 2016; 33(4):332–46.
2. Polat G. Advancing the multidimensional approach to family business professionalization. J Fam Bus Manag. 2020;
3. Kaplan RS, Robert NPDKS, Kaplan RS, Norton DP. The strategy-focused organization: How balanced scorecard companies thrive in the new business environment. Harvard Business Press; 2001.
4. Kaehr Serra C, Thiel J. Professionalizing entrepreneurial firms: Managing the challenges and outcomes of founder-CEO succession. Strateg Entrep J. 2019; 13(3):379–409.
5. Miller D, Friesen PH. Innovation in conservative and entrepreneurial firms: Two models of strategic momentum. Strateg Manag J [Internet]. 1982 Jan; 3(1):1–25. Available from: http://doi.wiley.com/10.1002/smj.4250030102
6. Martinez-Conesa I, Soto-Acosta P, Palacios-Manzano M. Corporate social responsibility and its effect on innovation and firm performance: An empirical research in SMEs. J Clean Prod. 2017; 142:2374–83.
7. Chang S, Shim J. When does transitioning from family to professional management improve firm performance? Strateg Manag J. 2015; 36(9):1297–316.
8. Madison K, Daspit JJ, Turner K, Kellermanns FW. Family firm human resource practices: Investigating the effects of professionalization and bifurcation bias on performance. J Bus Res. 2018; 84:327–36.
9. Dekker J, Lybaert N, Steijvers T, Depaire B. The effect of family business professionalization as a multidimensional construct on firm performance. J Small Bus Manag. 2015; 53(2):516–38.
10. Rosenbusch N, Brinckmann J, Bausch A. Is innovation always beneficial? A meta-analysis of the relationship between innovation and performance in SMEs. J Bus Ventur. 2011; 26(4):441–57.
11. Terzirovski M. Innovation practice and its performance implications in small and medium enterprises (SMEs) in the manufacturing sector: a resource-based view. Strateg Manag J. 2010; 31(8):892–902.
12. Méndez-Suárez M. Marketing Mix Modeling Using PLS-SEM, Bootstrapping the Model Coefficients. Mathematics [Internet]. 2021 Aug 3; 9(15):1832. Available from: https://www.mdpi.com/2227-7390/9/15/1832
13. Henseler J. Partial least squares path modeling: Quo vadis? Qual Quant. 2018; 52(1).
14. Jensen MC, Meckling WH. Theory of the firm: Managerial behavior, agency costs and ownership structure. J financ econ. 1976 Oct; 3(4):305–60.
15. Anderson RC, Reeb DM. Founding-family ownership and firm performance: evidence from the S&P 500. J Finance. 2003; 58(3):1301–28.
16. Lien Y-C, Li S. Professionalization of family business and performance effect. Fam Bus Rev. 2014; 27(4):346–64.
17. Chua JH, Chrisman JJ, Bergiel EB. An agency theoretic analysis of the professionalized family firm. Entrep theory Pract. 2009; 33(2):355–72.
18. Purkayastha S, Manolova TS, Edelman LF. Business group effects on the R&D intensity-internationalization relationship: Empirical evidence from India. J World Bus. 2018; 53(2):104–17.
19. Cavus MF, Demir Y. Institutionalization and corporate entrepreneurship in family firms. African J Bus Manag. 2011; 5(2):416.
20. Cherchem N. The relationship between organizational culture and entrepreneurial orientation in family firms: Does generational involvement matter? J Fam Bus Strateg. 2017; 8(2):87–98.
21. Gomez-Mejia LR, Makri M, Kintana ML. Diversification decisions in family-controlled firms. J Manag Stud. 2010; 47(2):223–52.
22. Gomez-Mejia LR, Cruz C, Berrone P, De Castro J. The bind that ties: Socioemotional wealth preservation in family firms. Acad Manag Ann. 2011; 5(1):653–707.
23. Morck R, Yeung B. Agency Problems in Large Family Business Groups. Entrep Theory Pract [Internet]. 2003 Jun 4 [cited 2021 Apr 24]; 27(4):367–82. Available from: http://journals.sagepub.com/doi/10.1111/1540-8520.101-1-00015
24. Lumpkin GT, Brigham KH, Moss TW. Long-term orientation: Implications for the entrepreneurial orientation and performance of family businesses. Entrep Reg Dev [Internet]. 2010 May [cited 2021 Apr 24]; 22(3–4):241–64. Available from: https://www.tandfonline.com/doi/abs/10.1080/08985621003726218
25. Putniņš TJ, Sauka A. Why does entrepreneurial orientation affect company performance? Strateg Entrep J [Internet]. 2020 Dec 27; 14(4):711–35. Available from: https://onlinelibrary.wiley.com/doi/10.1002/sej.1325
26. Sharpe WF. Capital asset prices: A theory of market equilibrium under conditions of risk. J Finance. 1964; 19(3):425–42.
27. Covin JG, Lumpkin GT. Entrepreneurial orientation theory and research: Reflections on a needed construct. Entrep theory Pract. 2011; 35(5):855–72.
28. Kafetzopoulos D. Performance management of SMEs: a systematic literature review for antecedents and moderators. Int J Product Perform Manag. 2020;
29. Hoskisson RE, Chirico F, Zyung J, Gambeta E. Managerial risk taking: A multitheoretical review and future research agenda. J Manage. 2017; 43(1):137–69.
30. Rafatnia AA, Suresh A, Ramakrishnan L, Abdullah DFB, Nodeh FM, Farajnezhad M. Financial Distress Prediction across Firms. J Environ Treat Tech. 2020; 8(2):646–51.
31. Anwar Z, Abbas K, Khan M, Razak DA. CSR DISCLOSURE AND FINANCIAL ACCESS: A CASE STUDY OF PAKISTAN. Int J Econ Manag Account [Internet]. 2019; 27(1):167–86. Available from: https://search.proquest.com/docview/2275868971?accountid=17225
32. Cannavale C, Nadali IZ. Entrepreneurial Orientations and Performance: A Problematic Explanatory Approach in the Iranian Knowledge-Based Industry. J Entrep [Internet], 2019 Mar 19 [cited 2021 Mar 24]; 28(1):86–93. Available from: http://journals.sagepub.com/doi/10.1177/0971355718810295
33. Dawson A. Private equity investment decisions in family firms: The role of human resources and agency costs. J Bus Ventur. 2011; 26(2):189–99.
34. Filatotchev I, Lien Y-C, Piesse J. Corporate governance and performance in publicly listed, family-controlled firms: Evidence from Taiwan. Asia Pacific J Manag. 2005; 22(3):257–83.
35. Habbershon TG, Williams ML. A resource-based framework for assessing the strategic advantages of family firms. Fam Bus Rev. 2012; 25(1):58–86.
36. Liang Q, Li X, Yang X, Lin D, Zheng D. How does family involvement affect innovation in China? Asia Pacific J Manag. 2013; 30(3):677–95.
37. Miller D, Le Breton-Miller I, Scholnick B. Stewardship vs. stagnation: An empirical comparison of small family and non-family businesses. J Manag Stud. 2008; 45(1):51–78.
38. Casillas JC, Moreno AM, Barbero JL. A configurational approach of the relationship between professionalization and growth of family firms. Fam Bus Rev. 2010; 23(1):27–44.
39. Classen N, Carree M, Van Gils A, Peters B. Innovation in family and non-family SMEs: an exploratory analysis. Small Bus Econ. 2014; 42(3):595–609.
40. Soto-Acosta P, Popa S, Palacios-Marqués D. E-business, organizational innovation and firm performance in manufacturing SMEs: an empirical study in Spain. Technol Econ Dev Econ. 2016; 22(6):885–904.
41. Byukusenge E, Munene J, Orobia L. Knowledge management and business performance: Mediating effect of innovation. J Bus Manag Sci. 2016; 4(4):82–92.
42. Gunday G, Ulusoy G, Kilic K, Alpkan L. Effects of innovation types on firm performance. Int J Prod Econ [Internet]. 2011 Oct; 133(2):662–76. Available from: https://linkinghub.elsevier.com/retrieve/pii/S0925527311002209
43. Artz KW, Norman PM, Hatfield DE, Cardinal LB. A Longitudinal Study of the Impact of R&D, Patents, and Product Innovation on Firm Performance. J Prod Innov Manag [Internet], 2010 Jul 7; 27(5):725–40. Available from: http://doi.wiley.com/10.1111/j.1540-5885.2010.00747.x
46. Jiménez-Jiménez D, Sanz-Valle R. Innovation, organizational learning, and performance. J Bus Res. 2011; 64(4):408–17.

47. Hajar I. The effect of business strategy on innovation and firm performance in the small industrial sector. Int J Eng Sci. 2015; 4(2):1–9.

48. Ndesaulwa AP, Kikula J. The impact of innovation on performance of small and medium enterprises (SMEs) in Tanzania: A review of empirical evidence. J Bus Manag Sci. 2016; 4(1):1–6.

49. González-Fernández M, González-Velasco C. Innovation and corporate performance in the Spanish regions. J Policy Model [Internet]. 2018 Sep; 40(5):998–1021. Available from: https://linkinghub.elsevier.com/retrieve/pii/S0161893818301054

50. Gallardo-Vázquez D, Valdez-Juárez LE, Castuera-Díaz AM. Corporate social responsibility as an antecedent of innovation, reputation, performance, and competitive success: A multiple mediation analysis. Sustainability. 2019; 11(20):5614.

51. Gallardo-Vázquez D, Barroso-Méndez MJ, Pajuelo-Moreno ML, Sánchez-Meca J. Corporate social responsibility disclosure and performance: A meta-analytic approach. Sustainability. 2019; 11(4):1115.

52. Darroch J. Knowledge management, innovation, and firm performance. J Knowl Manag. 2005;

53. Hashi I, Stošić N. The impact of innovation activities on firm performance using a multi-stage model: Evidence from the Community Innovation Survey 4. Res Policy. 2013 Mar; 42(2):353–66.

54. Atalay M, Anafarta N, Sarvan F. The Relationship between Innovation and Firm Performance: An Empirical Evidence from Turkish Automotive Supplier Industry. Procedia—Soc Behav Sci. 2013 Apr; 75:226–35.

55. Schumpeter JA. The Theory of Economic Development. Harvard University Press, Cambridge M, editor. 1934.

56. Wiklund J, Shepherd D. Knowledge-based resources, entrepreneurial orientation, and the performance of small and medium-sized businesses. Strateg Manag J. 2003; 24(13):1307–14.

57. Zhai Y-M, Sun W-Q, Tsai S-B, Wang Z, Zhao Y, Chen Q. An empirical study on entrepreneurial orientation, absorptive capacity, and SMEs’ innovation performance: A sustainable perspective. Sustainability. 2018; 10(2):314.

58. Mao CX, Zhang C. Managerial risk-taking incentive and firm innovation: Evidence from FAS 123R. J Financ Quant Anal. 2018; 53(2):867–98.

59. Merono-Cerdán AL, López-Nicolás C, Molina-Castillo FJ. Risk aversion, innovation and performance in family firms. Econ Innov new Technol. 2018; 27(2):189–203.

60. Kollmann T, Stöckmann C. Filling the entrepreneurial orientation–performance gap: The mediating effects of exploratory and exploitative innovations. Entrep Theory Pract. 2014; 38(5):1001–26.

61. Jeon I. The Impact of Entrepreneurship on Corporate Performance: Focusing on the Effects of Technological Innovation and Marketing Competence. Asia-Pacific J Bus Ventur Entrep. 2017; 12(3):87–105.

62. Faul F, Buchner A, Erdfelder E, Mayr S. A short tutorial of GPower. Tutor Quant Methods Psychol. 2007; 3(2):51–9.

63. Cohen J. Statistical power analysis for the behavioral sciences (2nd ed). Erbaum Press, Hillsdale, NJ, USA, 1988.

64. Van Gils A. Management and governance in Dutch SMEs. Eur Manag J. 2005; 23(5):583–9.

65. Fisher RJ. Social desirability bias and the validity of indirect questioning. J Consum Res. 1993; 20(2):303–15.

66. Hair JF, Ringle CM, Sarstedt M. Partial Least Squares Structural Equation Modeling: Rigorous Applications, Better Results and Higher Acceptance. Long Range Plann. 2013; 46(1–2):1–12.

67. Parada MJ, Nordqvist M, Gimeno A. Institutionalizing the family business: The role of professional associations in fostering a change of values. Fam Bus Rev. 2010; 23(4):355–72.

68. Damanpour F, Gopalakrishnan S. The Dynamics of the Adoption of Product and Process Innovations in Organizations. J Manag Stud [Internet]. 2001 Jan; 38(1):45–65. Available from: http://doi.wiley.com/10.1111/1467-6486.00227

69. Hughes A. Innovation and business performance: Small entrepreneurial firms in the UK and the EU. New Econ [Internet]. 2001 Sep; 8(3):157–63. Available from: http://doi.wiley.com/10.1111/1468-0041.00208

70. Uhlanaer LM, van Stel A, Duplat V, Zhou H. Disentangling the effects of organizational capabilities, innovation and firm size on SME sales growth. Small Bus Econ [Internet]. 2013 Oct 15; 41(3):581–607. Available from: http://link.springer.com/10.1007/s11187-012-9455-7

71. Ruiz-Palomó D, Diémez-Soto J, Duráñez A, Santos JAC. Family management and firm performance in family SMEs: The mediating roles of management control systems and technological innovation. Sustainability. 2019; 11(14).
72. Ringle CM, Wende S, Becker J-M. SmartPLS 3 [Internet]. Boenningstedt: SmartPLS GmbH; 2015. http://www.smartpls.com
73. Hair JF, Risher JJ, Sarstedt M, Ringle CM. When to use and how to report the results of PLS-SEM. Eur Bus Rev. 2019; 31(1):2–24.
74. Dijkstra TK, Henseler J. Consistent and asymptotically normal PLS estimators for linear structural equations. Comput Stat Data Anal. 2015; 81:10–23.
75. Henseler J. Bridging design and behavioral research with variance-based structural equation modeling. J Advert. 2017; 46(1):178–92.
76. Cepeda-Carrion G, Cegarra-Navarro JG, Cillo V. Tips to use partial least squares structural equation modelling (PLS-SEM) in knowledge management. J Knowl Manag. 2019; 23(1):67–89.
77. Aledo-Ruiz MD, Martinez-Caro E, Santos-Jaén JM. The influence of corporate social responsibility on students’ emotional appeal in the HEIs: The mediating effect of reputation and corporate image. Corp Soc Responsib Environ Manag [Internet]. 2021 Nov 23; Available from: https://onlinelibrary.wiley.com/doi/10.1002/csr.2221
78. Hair JF, Sarstedt M, Ringle CM, Gudergan SP. Advanced issues in partial least squares structural equation modeling. Sage Publications Sage CA: Los Angeles, CA; 2017.
79. Hair JF, Howard MC, Nitzl C. Assessing measurement model quality in PLS-SEM using confirmatory composite analysis. J Bus Res. 2020; 109(November 2019):101–10.
80. Henseler J, Hubona G, Ray PA. Using PLS path modeling in new technology research: Updated guidelines. Ind Manag Data Syst. 2016; 116(1):2–20.
81. Hair J, Hollingsworth CL, Randolph AB, Chong AYL. An updated and expanded assessment of PLS-SEM in information systems research. Ind Manag Data Syst. 2017; 117(3):442–58.
82. Yáñez-Araque B, Sánchez-Infante Hernández JP, Gutiérrez-Broncano S, Jiménez-Estévez P. Corporate social responsibility in micro-, small- and medium-sized enterprises: Multigroup analysis of family vs. nonfamily firms. J Bus Res. 2020;(October).
83. Chin WW, Dibbern J. Handbook of Partial Least Squares. Handb Partial Least Squares. 2010;171–93.
84. Voorhees CM, Brady MK, Calantone R, Ramirez E. Discriminant validity testing in marketing: an analysis, causes for concern, and proposed remedies. J Acad Mark Sci. 2016; 44(1):119–34.
85. del Valls Martínez M C, Martín-Cervantes PA, Sánchez Pérez AM, del Martínez Victoria M C. Learning Mathematics of Financial Operations during the COVID-19 Era: An Assessment with Partial Least Squares Structural Equation Modeling. Mathematics [Internet]. 2021 Sep 1; 9(17):2120. Available from: https://www.mdpi.com/2227-7390/9/17/2120
86. Dijkstra TK, Henseler J. Consistent Partial Least Squares Path Modeling. MIS Q. 2015; 39(2):297–316.
87. Hair JF, Hult GTM, Ringle C, Sarstedt M. A primer on partial least squares structural equation modeling (PLS-SEM). Sage publications; 2016.
88. Tenenhaus M, Vinzi VE, Chatelin YM, Lauro C. PLS path modeling. Comput Stat Data Anal. 2005; 48(1):159–205.
89. Khan GF, Sarstedt M, Shiaw WL, Hair JF, Ringle CM, Fritz MP. Methodological research on partial least squares structural equation modeling (PLS-SEM): An analysis based on social network approaches. Internet Res. 2019; 29(3):407–29.
90. Evermann J, Tate M. Assessing the predictive performance of structural equation model estimators. J Bus Res [Internet]. 2016; 69(10):4565–82. Available from: https://doi.org/10.1016/j.jbusres.2016.03.050
91. Hair JF, Sarstedt M. Factors versus Composites: Guidelines for Choosing the Right Structural Equation Modeling Method. Proj Manag J. 2019; 50(6):619–24.
92. Hair JF, Astrachan CB, Moisescu OI, Radomir L, Sarstedt M, Vaithilingam S, et al. Executing and interpreting applications of PLS-SEM: Updates for family business researchers. J Fam Bus Strateg. 2021 Sep; 12(3):100392.
93. Fanaz NA, Ahmed F, Ying M, Mehmood SA. The interplay of green servant leadership, self-efficacy, and intrinsic motivation in predicting employees’ pro-environmental behavior. Corp Soc Responsib Environ Manag [Internet]. 2021 Feb 15 [cited 2021 Mar 1];csr.2115. Available from: https://onlinelibrary.wiley.com/doi/10.1002/csr.2115
94. Falk RF, Miller NB. A primer for soft modeling. University of Akron Press; 1992.
95. Hair JF, Ringle CM, Gudergan SP, Fischer A, Nitzl C, Menictas C. Partial least squares structural equation modeling-based discrete choice modeling: an illustration in modeling retailer choice. Bus Res [Internet]. 2019; 12(1):115–42. Available from: https://doi.org/10.1007/s40685-018-0072-4
96. Chin WW. How to write up and report PLS analyses. In: Handbook of partial least squares. Springer; 2010. p. 655–90.
97. Hair JF Jr., Matthews LM, Matthews RL, Sarstedt M. PLS-SEM or CB-SEM: updated guidelines on which method to use. Int J Multivar Data Anal. 2017; 1(2):107.

98. Claver-Cortés E, Marco-Lajara B, Úbeda-García M, García-Lillo F, Rienda-García L, Zaragoza-Sáez PC, et al. Students' perception of CSR and its influence on business performance. A multiple mediation analysis. Bus Ethics A Eur Rev. 2020; 29(4):722–36.