FACTORS THAT CONTRIBUTE TO THE USE OF MODULARISATION IN THE AUTOMOTIVE INDUSTRY: A SURVEY IN BRAZIL

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ABSTRACT

This article aims to verify factors that can contribute to the use of modularisation in the Brazilian automotive industry through a survey methodology, using principal component analysis and structural equation modelling by partial least squares as techniques. We found three types of factors appearing as drivers or precursors of a formal strategy of modularisation: the use and knowledge of individual non-integrated aspects of modularisation, such as modular architecture; the introduction of competition differentials focusing mainly on cost reduction and increase in flexibility, quality, and dependability; and the existence of a business environment prone to innovation. The main conclusion is that elements of modularisation in the design and the need to meet customers’ requirements have much more influence in stimulating a formal, articulated modularisation strategy than a business environment prone to innovation.

INTRODUCTION

Continuous innovation processes allow the creation and maintenance of a competitive advantage in an industry, especially if the innovation processes embrace the entire life cycle of an industrial product [1-4]. In life cycle analysis, one of the main aspects is the architecture of the product [5-8]. Modular architecture products usually have longer life cycles than fixed architecture ones [9,10] as, with a modular architecture, companies do not need to develop new products for new demands [11,12]. The product acquires new functionalities by adding new modules and not by an entirely new development [11,13,14].

Researchers on innovation processes report that modularisation of products and processes can emerge due to innovation initiatives [15-18]. Studies emphasise that, when companies search for
innovative integrated solutions for products and processes, one of the alternatives is the use of modular architecture in the products [4,19,20]. They also point out that modularisation can accelerate product development processes [14,21,22] and increase manufacturability, which means designing products that are easier to manufacture [13,14,18,20]. Modularisation can also be helpful in meeting markets with characteristics of mass customisation [23-26].

A modularisation strategy can become an important part of a wider production strategy that relies on new products and innovative techniques to create competitive advantages in an industry [15,17,27-29]. Modularisation allows allying partners to develop independent modules individually, which reduces the need for formal knowledge transfer between companies [30]. At the same time, modularisation requires the formalisation of the interfaces, which stimulates some degree of functional dependence between partners [15] and generates specialised expertise in the members of the supply chain [31]. These arguments indicate the relevance of modularisation in the automotive industry. This article focuses on the relationship between modularisation, product design, competitiveness, and innovation in the Brazilian automotive industry. In 2016, 31 automakers and their respective supply chains operated in Brazil [32]. Two companies have operated in the modular concept since the 1990s: MAN, a truck manufacturer, and General Motors [33].

The main purpose of this article is to test factors that can contribute to the implementation of a modularisation strategy in the Brazilian automotive industry. The research question we pose is: What factors contribute most to the implementation of a strategy of modularisation in the Brazilian automotive industry? The research method is a survey. The specific objectives are to identify factors that may contribute to the formation of a modularisation strategy in the industry; to model and measure these factors in the Brazilian automotive industry; and to analyse the results according to multivariate statistical techniques.

2 REVIEW

Modularisation techniques are present in the design of new products, in manufacturing processes design [12,22], in organisational processes [27,34-36], and in service processes design [37-39]. Modularisation techniques make it easier to plan and control the degree to which changes in customer requirements affect the final product and the manufacturing process. Modularisation techniques also meet new demands faster, since they allow design decisions that are needed for the internal detailing of the parts [5,40] to be postponed until more information is available. Modularisation relies on modular architecture or integrated architecture, and has implications that can range from design engineering to business strategy [41]. The modular architecture includes designating specific functions to specific components of the physical structure of the product [42]. In a modular architecture, components communicate themselves using standardised interfaces. The integrated architecture includes the assigning of different functionalities to a single component. In this case, interfaces couple components, reducing the flexibility to recombine functions [43]. Modularisation allows the use of external sources of knowledge and innovation in product and process development processes, accelerating the solutions to meet customers’ demands [44]. Modularisation can be useful in developing products that are more competitive and in making manufacturing processes more efficient [40,45] while involving systemic decisions that reciprocally affect both the modules and the process design [5,46]. Modularisation can make manufacturing processes more efficient, as it allows a larger variety of products with fewer specific features to be produced at a reduced cost [14]. It can also increase flexibility in production and reduce product development and manufacturing lead-times [41,47]. Jacobs et al. [48] state that modularity in product design is the key factor in describing interrelationships and interactions between product and process modularity. Modularisation design elements can quickly create new products by recombining previously existing modules and processes [6].

Products with a modular architecture can increase flexibility, making manufacturing more responsive to customers’ demands [48]. Innovation initiatives can also stimulate modularisation [49]. Development processes relying on modularisation usually require companies to perform experiments based on new ideas. Such experiments can result in products and processes that are faster and cheaper since they usually start from an existing product or process [40]. Modular architectures also allow for independent innovations in modules, with no need for change in an entire product [15,19,22,29,40]. By contrast, radical innovations are more difficult to pursue due to the need to
abandon previous developments totally [50]. The insertion of products developed from modular
designs has contributed to the emergence of devices with universal compatibility in important
contemporary industries, such as the computer and communication devices industries [17,40].

In short, regarding the relationship with customers, modularity in product design can speed up new
product development, reduce development costs, and enhance customisation possibilities,
according to consumers' preferences. Regarding the relationship with the manufacturers, modularity
in process design can reduce setup times, reduce costs, and enhance revenues of low lot sizes
required for customised production [48].

Previous studies have observed that business environments that are prone to innovation have also
promoted the use of modularisation [15,17]. Specifically, in product design, modular structures are
favoured over integrated ones when rapid innovations are more important than performance [51].
Innovation is expected to use modularisation through two mechanisms: innovation within a
component or subsystem, and innovation by mix and match of existent modules [52]. The existence
of information systems, and the need to outsource activities can also facilitate the modularisation
of product design, and so reduce the scope of in-house activities. This kind of organisation requires
from the company a specific managerial organisation to comply with the formal requirements of a
modularisation strategy.

Finally, companies can respond to volatile markets simply by recombining existent resources and
modules, instead of creating entirely new products. The required agility for such recombination is
usually provided by a managerial and information structure that is suitable to handle the issues
related to modularity along the product life cycle [53]. In complex production networks, technical
changes in products strongly influence the managerial processes of companies and partners [54].
Usually, planning and controlling activities related to product changes require an appropriate
managerial structure [55,56].

3 THE RESEARCH

The research method is a survey [57]. Surveys are useful in research involving innovation, strategies,
products, and services [58,59]. The main research techniques are principal component analysis and
structural equation modelling by partial least squares (SEM-PLS) [60]. A SEM-PLS model has two
levels of analysis: the observation level and the theoretical level. On the observation level, the
outer model assesses latent variables, supported by indicators that capture manifest variables. On
the theoretical level, the inner model tests relationships among latent variables, which can be
endogenous or exogenous [61]. Latent variables connect each other via theoretical meanings,
expressed through hypotheses and supported by rationales [62].

Models can be reflective or formative. In reflective models, indicators of the same latent variable
are highly correlated, as they are different approaches to reflecting the essence of the same
variable. The latent variable causes the indicators, not the reverse, as in formative models [63].
Latent variables connect to manifest variables by factor loadings, which are inferred parameters
estimated from empirical associations among the observations of the manifest variables [62]. This
study uses only reflective models.

Previous studies help to derive the research hypotheses that support the initial inner model:

- **H1**: the use of modularisation issues (MOD, latent variable $\xi_1$) in design, even isolated,
  positively influences the creation of a modularisation strategy within the production strategy
  (MOD_STRT, latent variable $\xi_{10}$);
- **H2**: the potential of modularisation to create a competitive differential (DIF, latent variable
  $\xi_2$) positively influences the creation of a modularisation strategy within the production
  strategy (MOD_STRT, latent variable $\xi_{10}$); and
- **H3**: the existence of an environment prone to innovation in products and processes (INN, latent
  variable $\xi_3$) positively influences the creation of a modularisation strategy within the
  production strategy (MOD_STRT, latent variable $\xi_{10}$).
Latent variables (henceforward called ‘constructs’) are reflective, and integrate the inner model, in which three exogenous constructs ($\xi_1$, $\xi_2$, $\xi_3$) influence one endogenous construct ($\xi_{10}$). Figure 1 shows the initial inner model, based on the hypotheses.

![Figure 1: Initial inner model](image)

The survey included thirteen questions that investigated three exogenous constructs: modularisation issues (product architecture, effects on manufacturing, effects on mass customisation, and strategy support); competition (effects on cost, flexibility, delivery, and quality); and innovative environment in the company (business profile, new products, increase of revenue, customer satisfaction). The model also investigated the endogenous construct modularisation strategy in product design. For the survey, we used Google Forms [64]. The respondents used a Likert scale from 1 to 5, in which 1 represented ‘Fully disagree’ and 5 represented ‘Fully agree’, with intermediate degrees of agreement [65].

Previous studies must support the indicators and constructs of an initial theoretical outer model. Eventually, the initial outer model may change after the exploratory factor analysis (EFA). According to the EFA, the indicators may remain unchanged in the model, may exit, or may move among the constructs. Table 1 shows the initial outer model and the relevant literature supporting the manifest variables. All indicators will remain, but some will move after the EFA.

The research focus is on the Brazilian automotive industry. In 2017, 647 companies in the automotive industry (31 automakers and 616 auto parts manufacturers) and more than 5,500 dealers operated in Brazil. These companies spanned several business segments, such as automakers (automobiles, light commercial vehicles, trucks, buses, agricultural machinery, and self-propelled road machinery), and first- and second-tier auto parts suppliers, with revenues in 2015 of US$ 59.1 billion, making up 4.0 per cent of Brazil’s gross domestic product (GDP). The industry business yielded more than US$ 12 billion in taxes and employed more than 1.3 million people in 2015 [32].

Product development engineering managers, engineering directors, and designers answered the questionnaires (34 per cent are development engineers, 66 per cent are managers; 100 per cent are at least undergraduates, 69 per cent are graduates). The total number of valid responses is 117 companies: 19 automakers, 39 first-tier suppliers, and 59 second-tier suppliers. For reflective models, we used a rule of thumb presented by Barclay, Higgins, and Thompson [76]. The minimum sample size must be ten times the largest number of paths directed to a single construct. Since the maximum number of paths running into a single construct is four, the sample suffices. The rate of return was 50.43 per cent. The estimated rate of return is 31 per cent, and 46 per cent in this type of research [58]. Table 2 shows respondents’ data for the workforce and gross revenue in 2015, in US currency.
Table 1: Initial outer model

<table>
<thead>
<tr>
<th>Constructs</th>
<th>Indicators</th>
<th>Subjacent rationale</th>
<th>Authors</th>
</tr>
</thead>
<tbody>
<tr>
<td>Modularisation issues (MOD ≠1), exogenous</td>
<td>Use of modular architecture in the early stages of new product development (MOD1).</td>
<td>Importance of knowledge and application of modular architecture prior to the development of new products.</td>
<td>[5,15,19,20,29,42]</td>
</tr>
<tr>
<td></td>
<td>Use of integrated modularisation in the development of products and processes (MOD2).</td>
<td>Importance of the integrated application of modularisation in the development of products and processes.</td>
<td>[15-18,21,46,67]</td>
</tr>
<tr>
<td></td>
<td>Use of modularisation in meeting projects' requirements (MOD3).</td>
<td>Importance of having the necessary knowledge to find solutions based on modularisation.</td>
<td>[13,14,18,20,21,23,24,55,68]</td>
</tr>
<tr>
<td>Competitively (DIF ≠2), exogenous</td>
<td>Modularisation can reduce the cost of products and increase market share (DIF1).</td>
<td>Application of modularisation to create a competitive advantage by reducing costs and consequently increase market share.</td>
<td>[17,27,69,28,29,44,66]</td>
</tr>
<tr>
<td></td>
<td>Modularisation can help to meet mass customisation markets' demands (DIF2).</td>
<td>Application of modularisation to create a competitive advantage by flexibility and consequently increase market share in customised products.</td>
<td>[13,23-26]</td>
</tr>
<tr>
<td></td>
<td>Modularisation manufactures faster and reduces uncertainty in new products' development (DIF3).</td>
<td>Application of modularisation to create a competitive advantage by reducing average lead-time and variability in new product development and improve dependability.</td>
<td>[7,70,13,14,2,47,48,71]</td>
</tr>
<tr>
<td></td>
<td>Modularisation simplifies products, allowing part replacement (DIF4).</td>
<td>Application of modularisation to create a competitive advantage by improving quality and reliability indicators.</td>
<td>[5,7,15,16,19,27,29,50,56,67]</td>
</tr>
<tr>
<td>Environment prone to innovation (INN ≠3), exogenous</td>
<td>The company has an innovative management profile (INN1).</td>
<td>Perception of the degree to which the company stimulates innovation practices.</td>
<td>[4,66,73,74]</td>
</tr>
<tr>
<td></td>
<td>The company manages innovation to develop new products (INN2).</td>
<td>Perception of the degree to which the need for new products and processes causes the company to innovate.</td>
<td>[13,14,66,70]</td>
</tr>
<tr>
<td></td>
<td>Innovation contributes to increase revenue (INN3).</td>
<td>Perception of the degree to which new products and processes contribute to increasing revenue.</td>
<td>[66,73-75]</td>
</tr>
<tr>
<td></td>
<td>Innovation contributes to meeting customers' requirements (INN4).</td>
<td>Perception of the degree to which the company uses innovation to improve customers' satisfaction.</td>
<td>[15,16,19,27,28,50]</td>
</tr>
<tr>
<td></td>
<td>Existence of a specific, articulated strategy for modularisation activities (R1).</td>
<td>Extension and intensity of articulated actions promoting modularisation of products and services along the entire life cycle.</td>
<td>[15,17,27,69,28,29]</td>
</tr>
<tr>
<td>Modulatisation strategy within the production strategy (INN ≠10), endogenous</td>
<td>Existence of a managerial structure to handle modularisation (R2).</td>
<td>Efficiency and extension of the management activities regarding modularisation.</td>
<td>[12,53-56]</td>
</tr>
</tbody>
</table>
Table 2: Demographic information on the sample

<table>
<thead>
<tr>
<th>Class of company</th>
<th>Number of companies</th>
<th>Average revenue (Billions US$)</th>
<th>Average workforce</th>
</tr>
</thead>
<tbody>
<tr>
<td>Automakers</td>
<td>19</td>
<td>1,160</td>
<td>9,500</td>
</tr>
<tr>
<td>First-tier</td>
<td>39</td>
<td>185</td>
<td>900</td>
</tr>
<tr>
<td>Second-tier</td>
<td>59</td>
<td>76</td>
<td>430</td>
</tr>
</tbody>
</table>

4 RESULTS AND ANALYSIS

Supported by SPSS, we ran the EFA to check the latent structure of the indicators [62]. SPSS performs the Kaiser-Meyer-Olkin (KMO) and Bartlett sphericity tests to verify the adequacy of the sample. KMO produces a value between one and zero. Values above 0.60 suggest that the sample is appropriate [63]. The result was 0.922. Bartlett's sphericity test evaluates whether each sequential eigenvalue is significantly different from the remaining eigenvalues. Results should show p < 0.05 [77]. The result was p = 0.00. Previously reviewed studies provided the indicators of the outer model.

Some of the indicators distributed according to unexpected factors, so we reorganised the constructs [63]. The best-achieved reorganisation relies on the extraction of three factors and the application of the Equamax rotation and Kaiser normalisation. The explained variance is 75.5 per cent of the total variance. Explained variance must exceed 70 per cent [77]. The reorganisation of the constructs and the factors’ extraction assured unidimensionality. All loads exceed 0.6 [78]. Therefore all indicators remained in the model. Table 3 shows the EFA result, the old and new acronyms, and keywords to the indicators, which help to understand the new nomenclature of the constructs. Values in bold refer to the same construct.

Table 3: EFA

<table>
<thead>
<tr>
<th>Old acronym</th>
<th>New acronym</th>
<th>Factors</th>
<th>Keywords</th>
</tr>
</thead>
<tbody>
<tr>
<td>MOD1</td>
<td>Design1</td>
<td>1</td>
<td>0.17 0.78 0.35 early stages of development</td>
</tr>
<tr>
<td>MOD2</td>
<td>Design2</td>
<td>2</td>
<td>0.23 0.64 0.56 integrated modularisation</td>
</tr>
<tr>
<td>MOD3</td>
<td>Customer1</td>
<td>3</td>
<td>0.35 0.56 0.60 projects’ requirements</td>
</tr>
<tr>
<td>DIF1</td>
<td>Design3</td>
<td>1</td>
<td>0.07 0.64 0.51 increases market share</td>
</tr>
<tr>
<td>DIF2</td>
<td>Design4</td>
<td>2</td>
<td>0.39 0.74 0.17 mass customisation</td>
</tr>
<tr>
<td>DIF3</td>
<td>Customer2</td>
<td>3</td>
<td>0.26 0.30 0.78 reduces uncertainty</td>
</tr>
<tr>
<td>DIF4</td>
<td>Customer3</td>
<td>1</td>
<td>0.25 0.27 0.80 simplifies products</td>
</tr>
<tr>
<td>INN1</td>
<td>Innov1</td>
<td>2</td>
<td>0.76 0.05 0.34 innovative profile</td>
</tr>
<tr>
<td>INN2</td>
<td>Innov2</td>
<td>3</td>
<td>0.81 0.39 0.08 innovation and new products</td>
</tr>
<tr>
<td>INN3</td>
<td>Innov3</td>
<td>1</td>
<td>0.82 0.15 0.24 innovation and revenue</td>
</tr>
<tr>
<td>INN4</td>
<td>Innov4</td>
<td>2</td>
<td>0.78 0.31 0.24 innovation and customers</td>
</tr>
</tbody>
</table>

The three endogenous constructs are design, customers, and innovation.

The ‘design’ construct refers to the recognition that the company uses modularisation techniques in design. The subjacent rationale [62] of the construct is: the application of modularisation in the early stages of development, the use of integrated modularisation in the products and manufacturing processes, reduced costs and increased market-shares achieved by modularisation, and mass customisation requirements fulfilled by modularisation, indicating that the company uses modularisation in the design phase of new product development.
The ‘customers’ construct refers to the recognition that the company uses modularisation techniques to meet customers’ requirements. The subjacent rationale of the construct is: the application of modularisation to fulfill projects’ requirements, to reduce uncertainty, and to simplify products, indicating that the company uses modularisation to meet customers’ requirements.

The ‘innovation’ construct refers to the recognition of a consistently innovative behaviour. The subjacent rationale of the construct is: an innovative managerial profile, the use of innovative techniques in new products, the existence of revenues due to innovation, and the use of innovation to meet customers’ requirements, indicating that the company has a consistent innovative behaviour.

The exogenous construct is ‘modularisation strategy’. The rationale is: a specific articulated strategy to deal with modularisation techniques and a managerial structure to handle modularisation activities and their implications, indicating that a company has a modularisation strategy inserted in its operation and production strategy.

Finally, the rationale for the inner model is: if a company uses modularisation in design, uses modularisation in meeting customers’ requirements, and has a consistent innovative behaviour, then this company also has a modularisation strategy inserted in its operation and production strategy.

Figure 2 shows the model and the results of the PLS algorithm, calculated by the software SmartPLS 2.0 M3 [79].

Hair et al. [63] recommend assuring internal consistency reliability, convergent validity, and discriminant validity for a reflective model. For the structural model, they recommend verifying the coefficient of determination, the size and significance of path coefficients, the effect size f² in the coefficient of determination, and the cross-validated redundancy Q2.

Cronbach’s alphas and composite reliabilities can help to verify the internal consistency reliability. The Cronbach’s alphas and composite reliabilities for [ξ1, ξ2, ξ3, ξ10] are [0.86, 0.84, 0.88, 0.68] and [0.90, 0.90, 0.89, 0.86] respectively. Alphas must exceed 0.6. Composite reliabilities must range from 0.6 to 0.9 in exploratory studies [63].

Average extracted variances (AVE) and outer loadings or indicator reliabilities (the indicator reliability is the square of the outer loading) can verify the convergent validity. The AVEs for [ξ1,
AVEs must be larger than 0.5 to confirm convergent validity [78]. All outer loadings are higher than 0.708 (all indicator reliabilities are higher than 0.50). Bagozzi, Yi, and Phillips [80] recommend retaining unconditionally indicators with outer loadings higher than 0.708 (reliabilities higher than 0.5), removing unconditionally indicators with outer loadings lower than 0.4 (reliabilities lower than 0.16), and considering the differences in reliabilities and validities before and after removing indicators with outer loadings in the interval [0.4 - 0.708].

Cross-loadings and AVE can verify the discriminant validity. The reallocation of indicators favoured discriminant validity [81]. For all indicators, the outer loading on the own construct is greater than any cross-loadings, i.e., the loading of the indicator on the other constructs. For discriminant validity, the square root of the AVE should be higher than any correlation among the constructs [82]. Table 4 shows the application of the Fornell-Larcker criterion. In the diagonal, in bold, the table shows the square root of the AVEs.

Table 4: Discriminant validity test

<table>
<thead>
<tr>
<th></th>
<th>ξ1</th>
<th>ξ2</th>
<th>ξ3</th>
<th>ξ10</th>
</tr>
</thead>
<tbody>
<tr>
<td>ξ1</td>
<td>0.84</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>ξ2</td>
<td>0.83</td>
<td>0.87</td>
<td></td>
<td></td>
</tr>
<tr>
<td>ξ3</td>
<td>0.62</td>
<td>0.58</td>
<td>0.85</td>
<td></td>
</tr>
<tr>
<td>ξ10</td>
<td>0.75</td>
<td>0.74</td>
<td>0.59</td>
<td>0.87</td>
</tr>
</tbody>
</table>

For the structural model, the criteria used are the coefficient of determination R2 (the percentage of variance in the endogenous construct explained by the exogenous constructs), path coefficients, effect size f2, and cross-validated redundancy Q2 [78].

R2 is 0.792. As it exceeds 60 per cent, the structural model is consistent [63]. The path coefficients for [ξ1 → ξ10 ; ξ2 → ξ10; ξ3 → ξ10] are [0.352, 0.344, 0.172]. Path coefficients have standardised values ranging from -1 to +1; the closer to 1 or -1, the stronger the positive or negative relationship. One method of assessing the significance of the path coefficients is by t-values and significance, using the bootstrapping method. Table 5 shows the significance of the path coefficients for p-values of 1 per cent and 5 per cent.

Table 5: Significance of path coefficients

<table>
<thead>
<tr>
<th>path</th>
<th>Size of the coefficient</th>
<th>t statistics</th>
<th>Significance (p&lt;0.05, one-tailed, critical level = 1.65)</th>
<th>Significance (p&lt;0.01, one-tailed, critical level = 2.33)</th>
</tr>
</thead>
<tbody>
<tr>
<td>ξ1 → ξ10</td>
<td>0.352</td>
<td>2.36</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>ξ2 → ξ10</td>
<td>0.344</td>
<td>2.30</td>
<td>Yes</td>
<td>No</td>
</tr>
<tr>
<td>ξ3 → ξ10</td>
<td>0.172</td>
<td>1.75</td>
<td>Yes</td>
<td>No</td>
</tr>
</tbody>
</table>

The impact that a single exogenous construct has on an endogenous one is the f2 effect. To calculate the effect of an exogenous construct, the SEM-PLS algorithm must run twice, with all constructs and without the specific constructs. The larger the difference in R2, the more influence the construct has. Cohen [83] states that values around 0.35, 0.15, and 0.02 respectively represent large, medium, and small effect sizes. Equation 1 provides the f2 effect size [63]. Table 6 shows the f2 effect.

\[
f^2 = \frac{R^2_{included} - R^2_{excluded}}{1 - R^2_{included}}
\] (1)
Finally, the Q2 value indicates the model’s predictive relevance for an endogenous construct—that is, the model’s out-of-the-sample predictive power. A Q2 larger than zero for an endogenous reflective construct means that the model exhibits predictive relevance. Otherwise, it lacks predictive relevance. The calculation of Q2 requires the blindfolding procedure. As with \( f^2 \), the \( q^2 \) effect size is calculated by omitting the constructs one by one. The resulting \( q^2 \) effect size serves to classify the relevance of the endogenous construct. Calculated with an equation similar to Equation 1, values around 0.02, 0.15, and 0.35 imply small, medium, or large predictive relevance [78]. Table 7 shows the \( q^2 \) effect in the measurement model.

### Table 7: \( q^2 \) effect size

<table>
<thead>
<tr>
<th>Structural model</th>
<th>Q2</th>
<th>( q^2 )</th>
<th>Effect size</th>
</tr>
</thead>
<tbody>
<tr>
<td>All ( \xi ) included</td>
<td>0.451</td>
<td>0.064</td>
<td>Medium</td>
</tr>
<tr>
<td>without ( \xi_1 )</td>
<td>0.416</td>
<td>0.040</td>
<td>Small</td>
</tr>
<tr>
<td>without ( \xi_2 )</td>
<td>0.429</td>
<td>0.056</td>
<td>Medium</td>
</tr>
</tbody>
</table>

Combining the various criteria, we conclude heuristically that all constructs are relevant, but that \( \xi_1 \) and \( \xi_2 \) are more relevant than \( \xi_3 \). The relevance means that, in Brazil’s automotive industries, companies are more prone to formalise a modularisation strategy if they already have some initiatives in the design phase and if they are concerned about creating a competitive differential, than if they simply have an innovative business environment.

## 5 CONCLUSION

Our study concludes that, in the Brazilian automotive industry, all the studied factors contribute to the implementation of a modularisation strategy. However, the degree of contribution varies according to the factor. Variables connected to the design of the product are more influential than variables connected to the fulfilment of customers’ requirements, and both are more influential than innovation. Albeit to a lesser extent, the pre-existence of a business environment that favours innovation should also be considered in further studies.

This study has implications for engineering management and product strategy practices, especially in the automotive industry. If managers want to create or increase the use of modularisation in the industry, they have two good options. The first is to use modularisation techniques, even isolated or non-articulated, with process modularisation from the initial phase—the design of the product. The second is to try to fulfil customers’ requirements with the use of modularisation techniques, instead of other methods. As the literature reports, the inclusion of modular architecture in the product is relevant to promote an articulated strategy that includes a systemic adoption of modularisation. It is also remarkable that the average scores (retrieved from the questionnaires) of the constructs ‘design’ and ‘customer’ (which have more influence) do not differ much (3.9 and 3.6 respectively), but both are less than the score of innovation (4.08), which has less influence. The difference indicates that the hierarchy of intensities and capacities to influence are different, meaning that, in a general sense, the industry does not pursue a modularisation strategy—or, at least, does not have a roadmap to do it. Further research using different methodologies such as action research is necessary to confirm a road map to modularisation.
For further research, we suggest a confirmatory replication. Eventually, new studies should consider the inclusion of a new construct for correlated technologies, such as flexible manufacturing and automation. Finally, we suggest the replication with the same sample over time, to show the evolution of the factors in the Brazilian automotive industry.

REFERENCES


