A Vector Error Correction Model to Estimate Potato Supply Response in Colombia

Un modelo de corrección de errores vectoriales para estimar la respuesta de la oferta de papa en Colombia

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Abstract: The present study aims to investigate the relationship between potato output supply and potato and fertilizer prices in Colombia using 2003-2018 time series data. Johansen’s cointegration and the vector error correction framework were employed, and Granger and diagnostic tests were performed. The empirical results show a strong causal relationship between the variables and that the signs and the magnitude of coefficients are statistically significant. Potato and fertilizer prices are inelastic, and the speed adjustment coefficient of 70% implies that it takes a short time for farmers to respond to price movements. It can be concluded that own-price and fertilizer prices are significant in determining output supply responsiveness in both the short and long run.

Keywords: Agricultural supply response, potato prices, cointegration, vector error correction

Resumen: El objetivo del presente estudio fue investigar la relación entre la producción de papa y los precios de la papa y de los fertilizantes en Colombia utilizando datos de series temporales del 2003-2018. Se utilizó el análisis de cointegración de Johansen y el modelo de corrección de errores y se realizaron pruebas de diagnóstico y de Granger. Los resultados empíricos mostraron que los signos y la magnitud de los coeficientes fueron estadísticamente significativos y que había una fuerte relación causal entre las variables. Los precios de la papa y de los fertilizantes fueron inelásticos y el coeficiente de ajuste de velocidad del 70% implica que los agricultores tardan poco en responder a los movimientos de los precios. Se puede concluir que los precios de la papa y de los fertilizantes fueron significativos para determinar la capacidad de respuesta de la oferta de papa a corto y largo plazo.

Palabras clave: oferta, precio de la papa, cointegración, corrección de errores
Introduction

The potato, *Solanum tuberosum* L. (Solanaceae), is native to the Andes Mountains in South America, ranks as the world’s third most important food crop after wheat and rice, and is grown in over 125 countries (Dlamini, 2018; Zhang et al., 2017). China is currently the largest potato producer at 99,205,580 tons (t) per year, followed by India (48,605,000 t) and Russia (29,589,976 t) (Food and Agriculture Organization of the United Nations [FAOSTAT], 2020; Ministerio de Agricultura y Desarrollo Rural [MADR], 2020).

In Colombia, potatoes are grown at altitudes of 2,200 to 3,400 meters above sea level, and there are more than thirty known varieties, of which only ten are considered of commercial importance (Agronet, 2020; Osorio et al., 2012). Potato production in 2019 was reported at 2,701,062 t, with a total cultivated area of 128,622 hectares (ha) and an average yield of 21.0 t/ha. Potato production was concentrated in the departments of Cundinamarca (37 %), Boyacá (27 %), Nariño (20 %) and Antioquia (6 %) (Agronet, 2020).

Potato production is undoubtedly a vital part of the total Colombian economy, as it accounts for 3.3 % of the country’s Agricultural Gross Domestic Product and is the primary source of income for more than 100,000 farmers. During the 2015-2018 period, potato production was responsible for creating 264,000 jobs, of which 75,000 were for directly employed workers and 189,000 for indirectly employed workers (Federación Colombiana de Productores de Papa [Fedepapa], 2020; MADR, 2019).

The study of agricultural supply response has long been a subject of discussion among economists and has received much attention in the literature. Various methods have been used to study agricultural supply response, the appropriateness of which depends to some extent on the type of crop, the available data, and the decision-making processes they are intended to support. According to Huq and Arshad (2010), most empirical research is based on the application, modification, and extension of the Nerlove (1958) framework. The Nerlove model, which is a partial adjustment and adaptive expectations model, describes the dynamics of supply response using price expectations. In its traditional version, it consists of three equations where the optimal output ($Q^*_t$) is a function of expected price ($P^e_t$), as shown in equation (1) (Askari & Cummings, 1977; Kohli, 1996).

$$Q^*_t = a + bP^e_t$$  \(1\)

Furthermore, it is assumed that price expectations are adaptive and, as such, are reviewed every year and adjusted to the difference between the actual price in the previous period and the expected price. A delayed output adjustment within the agricultural production cycles is also assumed, as shown in equations (2) and (3).

$$P_t - P^e_{t-1} = \gamma (P_{t-1} - P^e_{t-1}) \quad 0 \leq \gamma < 1$$  \(2\)

$$Q_t - Q_{t-1} = \delta (Q^*_t - Q_{t-1}) \quad 0 \leq \delta < 1$$  \(3\)
The Nerlovian approach has several advantages, including that it is a simple procedure and allows flexibility when developing dynamic specifications (Shahzad et al., 2018). Nevertheless, the major criticisms of the approach are that the formal theory it uses is inadequate and that when using the standard ordinary least square (OLS) technique, spurious regression results may occur if variables are non-stationary. Moreover, using conventional empirical procedures means that own-price short- and long-run supply responses are likely to be underestimated (Obayelu & Ebute, 2016; Tripathi & Prasad, 2009).

Another commonly employed model for estimating supply response is the dual approach derived from the primal profit maximization and cost minimization methods. It assumes that farmers are profit-maximizing agents, price takers, and technically efficient and originates from a study by Hotelling (1932). From an empirical point of view, it requires detailed input prices, and the profit and demand function equations should be simultaneously estimated (Arnade & Kelch, 2007). According to Khan et al. (2018), the dual approach has been widely used in applied economics and has proven useful when analyzing the impact of price policy changes on farmers’ behavior. However, its major drawbacks are that the model uses integrated (non-stationary) series, which can lead to spurious regression, and is static. In other words, it does not present specific time lags; therefore, it is difficult to determine whether the estimated elasticities are short- or long-run values (Henneberry & Tweeten, 1991; Kohli, 1996; Tripathi & Prasad, 2009).

Given the limitations of the above supply response approaches, cointegration analysis using the error correction model (ECM) is believed to be a viable alternative, as it captures both the short-run dynamic relationship between cointegrated variables and their adjustments toward long-term equilibrium (Obayelu & Ebute, 2016). Furthermore, it provides both short and long-run elasticities and is consistent in dealing with non-stationary data (Lutkepohl, 2005). Two well-known and widely used methods for verifying cointegration are the Engle-Granger (1987) and Johansen (1988) procedures.

The Engle-Granger two-step method is based on a single equation that can be estimated using OLS regression and applying unit root tests to the residuals to identify stationarity. Johansen’s method, however, does not rely on OLS and instead uses a maximum likelihood estimation technique. Its advantage over the Engle-Granger procedure is that it can test two or more cointegrated series. On the other hand, one of its main weaknesses is that it requires a high number of observations, which are sometimes unavailable (Pekmezci & Dilek, 2014).

Compared with the number of studies that have used variants of the Nerlovian approach, fewer studies have used the vector error correction model (VECM) and cointegration analysis to examine supply response to price. However, Huq and Arshad (2010) have used the VECM approach to estimate the supply response of potato farmers in Bangladesh for the 1982-2006 period. Their study found that producers react to harvest prices, as noted by significant short- and long-run elasticities. Likewise, Obayelu and Ebute (2016) have used cointegration and VECM to investigate the responsiveness of cassava output supply to price changes in Nigeria in the 1966-2010 period. Their results show a linear deterministic trend in the data and that cultivated area and own-prices jointly explain the variation in cassava output in both the short and long run.
Obayelu and Salau (2010) estimated the impact of Nigerian agricultural supply on prices and exchange rates in the 1970-2007 period. They concluded that both variables were significant in determining supply in the short and long run. Similarly, Mesike et al. (2010) evaluated the supply response of rubber farmers in Nigeria from 1970 to 2008 using error correction techniques and found that producer prices and a policy shift in the form of structural adjustment programs significantly affected the supply of rubber.

Mose et al. (2007) estimated the supply response of maize farmers in the Trans-Nzoia District for the 1980-2002 period. Their study found that farmers responded strongly to price incentives, as shown by the significant but negative short- and long-run elasticity of supply to fertilizer price and positive own-price elasticity in the short and long run.

According to the studies described above, potato supply is highly influenced by market price: price increases tend to increase output, and vice-versa. In addition, the time lag in terms of production is an important issue, as output is obtained months after planting; consequently, changes in market conditions raise many challenges for farmers (Huq & Arshad, 2010). Understanding the output supply response to price changes proves not only to be helpful information for evaluating the effectiveness of price policies in farmers’ resource allocation but also facilitates farmers’ output decisions. Price policies have long been the basis of farmers’ decisions in many countries, and any incentives or production planning will require detailed knowledge of supply factors (Tripathi & Prasad, 2009).

Given these points, the underlying aim of the present study is to estimate the potato output supply response to own-price and fertilizer price. Hence, based on Johansen’s cointegration analysis, a VECM was used as the appropriate tool for price transmission analysis. It not only overcomes the problems of spurious regression but also provides information about the short- and long-term relationships between time series.

The present paper is structured as follows: after the introduction, the second section provides relevant data and describes the proposed econometric methodology, the third section includes the empirical results, and the final two sections discuss the results and summarize the significant findings of this study.

Materials and methods

An explanatory, correlational, and non-experimental research design was selected to describe the relationships between the variables. The information required for the present study was obtained from the FAOSTAT, Fedepapa, and MADR databases.

Data
The data used in this study consist of annual time series spanning from 2003 to 2018. Due to the limited scope of this study, the variables were restricted to the mean annual potato and fertilizer prices in Colombian pesos (COP) and the country’s total potato production (tons).
The potato prices summarized in table 1 were obtained from FAOSTAT and deflated using the 2018 Colombian producers’ price index (IPP). All series were converted to natural logarithm form for economic interpretation. The data and regression model were analyzed using the E-views® 11 package (E-views, 2020).

Table 1. Potato prices 2003-2018

| Year | COP/ton |
|------|---------|
| 2003 | 1,260,905 |
| 2004 | 1,159,956 |
| 2005 | 1,339,083 |
| 2006 | 1,378,088 |
| 2007 | 713,034  |
| 2008 | 1,240,995 |
| 2009 | 1,250,109 |
| 2010 | 1,385,438 |
| 2011 | 1,185,593 |
| 2012 | 782,427  |
| 2013 | 578,830  |
| 2014 | 941,342  |
| 2015 | 692,435  |
| 2016 | 832,827  |
| 2017 | 849,327  |
| 2018 | 857,048  |

Source: FAOSTAT

Empirical specification

In a vector autoregression (VAR) model, all variables are considered endogenous, with each variable written as a regression of past lags of itself and the past lags of all other variables in a finite order system (Lawrence, 2012). When the variables are integrated as order I (1) and there is cointegration (i.e., there are relationships amongst the variables), the VECM can be derived from the vector autoregressive model (VARM); hence, the VEC model is a VAR model with cointegration constraints. However, a preliminary analysis of the data must be performed to confirm that the model specification is correct; this can consist of the unit root (stationarity) and cointegration tests. According to Hamilton (1994) and Lutkepohl (2005), the VEC model can be parameterized according to equation (4):

\[
\Delta Y_t = \alpha \beta Y_{t-1} + \sum_{i=1}^{p-1} \Gamma_i \Delta Y_{t-i} + \varepsilon_t
\]  

(4)

Where:

- \(\Delta Y_t = Y_t - Y_{t-1}\); \(Y_t\) is a p-dimensional column vector of non-stationary I (1) endogenous variables.
- \(\beta Y_{t-1}\) is the error correction term that reflects long-term equilibrium relationships between variables.
- \(\alpha\) is a full rank matrix that determines the speed of adjustment towards equilibrium.
- \(\Gamma\) is a \(n \times 1\) vector-matrix representing the lagged variables’ coefficients.
Thus, equation (5) was proposed:

\[
\ln Q_t = \varphi_0 + \varphi_1 \ln Q_{t-1} + \varphi_2 \ln PP_{t-1} - \varphi_3 \ln PF_{t-1} + \varepsilon_t
\]  

(5)

Where:

- \(\ln Q_t\) is the quantity supply,
- \(\ln Q_{t-1}\), \(\ln PP_{t-1}\), and \(\ln PF_{t-1}\) are the lagged quantity and own potato and fertilizer prices;
- \(\varphi_0\) is the constant or intercept;
- \(\varphi_1\), \(\varphi_2\), and \(\varphi_3\) are the long-run coefficients, and \(\varepsilon_t\) is the error term.

The sign of each coefficient indicates the direction of the relationship between the independent and the response variables. A positive sign for \(\ln PP\) is expected, as it indicates a direct relationship with production level. Meanwhile, \(\ln PF\) should present an inverse relationship, as an increase in fertilizer prices implies higher production costs for farmers, thus reducing output supply.

**Stationarity, cointegration, and validity tests**

Testing for stationarity and cointegration in autoregressive processes is key to identifying the econometric model that best fits the times series; therefore, the Augmented Dickey-Fuller (ADF) (1979) and Johansen’s (1988) cointegration tests were performed. In addition, model validation was carried out to verify that the residuals met the assumptions of stability, normality, constant error variance, and unrelated error terms.

**Granger causality test**

The Granger (1969) causality test was applied to determine the relationship between the time series. It assumes that if Z helps predict W, the regression of W is based on its past values. Past values of Z are added, and therefore Z can be called the Granger cause of W; otherwise, it is called the non-Granger cause.

**Results**

Unit root tests were conducted as a first step towards specifying the correct autoregressive model. The results of the ADF test are summarized in table 2 for level and first difference with intercept and trend and intercept. The calculated ADF statistics for all log-level series with intercept, trend, and intercept were lower in absolute value than their respective critical values at 1 %, 5 %, and 10 % significance levels, which implies that they were not stationary; as such, the null hypothesis of a unit root cannot be rejected. However, at the first difference, all log series values were higher at 5% and 10% significance levels; therefore, they do not show unit root and are integrated in order I (1).
Table 2. Stationarity of series ADF test

| Variables          | ADF test value | Critical value |
|--------------------|----------------|----------------|
|                    |                | 1%  | 5%  | 10% |
| Level              |                |     |     |     |
| Intercept          | -2.473         | -3.959 | -3.081 | -2.681 |
| lnQ                | -1.310         | -4.004 | -3.098 | -2.690 |
| lnPP               | -1.878         | -4.004 | -3.098 | -2.690 |
| Trend and intercept|                |     |     |     |
| lnQ                | -3.241         | -4.800 | -3.791 | -3.342 |
| lnPP               | -2.020         | -3.959 | -3.081 | -2.681 |
| lnPF               | -3.198         | -4.9922 | -3.875 | -3.388 |
| First difference   |                |     |     |     |
| Intercept          | -3.475         | -4.004 | -3.099 | -2.690 |
| lnQ                | -4.715         | -4.004 | -3.098 | -2.690 |
| lnPP               | -5.720         | -4.004 | -3.098 | -2.690 |
| Trend and intercept|                |     |     |     |
| lnQ                | -3.806         | -4.004 | -3.099 | -2.690 |
| lnPP               | -6.744         | -4.004 | -3.098 | -2.690 |
| lnPF               | -5.407         | -4.004 | -3.098 | -2.690 |

Source: Elaborated by the authors

For the cointegration analysis, Johansen’s Maximum Likelihood method was employed, and the results show that, in both the trace and maximum Eigenvalue tests, the statistical values for the null hypothesis of no cointegrated equations were greater than their critical values, with a probability below the 5% significance level (table 3). The preceding confirms the existence of one cointegrated equation and hence a long-term relationship between potato supply and potato and fertilizer prices, and therefore VEC modeling could continue to be conducted.

Table 3. Johansen’s (1988) cointegration test

| Null hypothesis | Eigenvalue | Trace statistic | Critical value (0.05) | p-value | Maxeigen statistic | Critical value (0.05) | p-value |
|-----------------|------------|-----------------|----------------------|---------|--------------------|----------------------|---------|
| None            | 0.923      | 50.9590         | 29.7870              | 0.0001  | 36.0117            | 21.1316              | 0.0002  |
| At most 1       | 0.606      | 14.9470         | 15.4940              | 0.0603  | 13.0688            | 14.2646              | 0.0766  |
| At most 2       | 0.1255     | 1.8789          | 3.8140               | 0.1704  | 1.8789             | 3.84146              | 0.1704  |

Source: Elaborated by the authors

**VECM estimation**

The error correction model presented in table 4 captures the long- and short-run relationships between variables in the integrated series. However, lag length criteria were established prior to the model’s estimation, as incorrect fitting can lead to autocorrelated errors and forecast miscalculations. Both the Akaike information criteria and the VEC lag order specification (p.1) suggested a lag length of one.
Table 4. Long- and short-run VECM estimates

| Cointegrating equation | Coefficient | t-value |
|------------------------|-------------|---------|
| lnQ_{t-1}              | 1.00        |         |
| lnPP_{t-1}             | 0.369*      | 7.096   |
| lnPF_{t-1}             | -1.394*     | -5.383  |
| C                      | 3.6478*     |         |

| Error correction      | ΔLQ         | t-value |
|-----------------------|-------------|---------|
| ECt_{t-1}             | -0.702*     | -2.979  |
| ΔlnQ_{t-1}            | -0.918*     | -3.342  |
| ΔlnPP_{t-1}           | 0.831*      | 2.483   |
| ΔlnPF_{t-1}           | -0.496*     | -3.868  |
| C                     | 0.005       | 0.224   |
| R-squared             | 0.724       |         |
| F-statistic           | 5.929       |         |

*Statistical significance level at 5%

Source: Elaborated by the authors

The results show that the signs and the magnitude of the estimated supply response coefficients were within a reasonable range and that the model fits the observed data, as indicated by the R² (0.74) and F (5.92) statistics (table 3). As all variables were in logarithmic form, the estimated cointegration coefficients represent supply elasticities. Therefore both the short- (0.83) and long-run (0.36) own-price coefficients were positive, significant (p < 0.05), and inelastic, suggesting a positive relationship with potato supply and that a percentage increase in lagged potato price leads to a less than proportionate increase in supply.

As expected and in line with the theory, the estimated fertilizer price coefficients presented negative elasticity in the short (-0.49) and long-run (-1.39), which indicates that as the price of an input increases, the costs increase, making farmers reduce output.

Additionally, given that potato crops have high fertilizer requirements, more remarkable changes in supply response to price movements this is expected. Moreover, considering the structure of potato production, the negative sign on the lagged output suggests that potato crops could be replaced with substitute crops in the event of excess supply. In other words, an increase in potato output would result in a price decrease and induce farmers to shift to other crops. Accordingly, the error correction coefficient of the short-run equation, which determines the speed of adjustment towards long-run equilibrium, presented the correct negative sign (-0.70) and was statistically significant (p < 0.05), implying that the rate at which potato and fertilizer prices adjust to long-run equilibrium is 70% within one year.

**Granger causality test**

This study also examines causal relationships, which can provide helpful information about the model’s ability to make predictions. The results show a robust causal relationship, as the t-
statistics of the independent variables and the error correction term are statistically significant ($p < 0.05$) (table 5). In addition, if the $p$-values of the log time series obtained from the Granger test were lower than the significance level of 5%, the null hypothesis could be rejected, thus confirming the existence of causality among the variables. Except that $\ln Q$ and $\ln PP$ are not Granger causes of $\ln PF$, all other variables present Granger causality relationships: $\ln PP$ and $\ln Q$ are Granger causes of each other and $\ln PF$ is the Granger cause of both $\ln Q$ and $\ln PP$. Reciprocity among the three variables was established (table 5).

### Table 5. Granger causality test

| Dependent variable D(LQ) | Excluded  | Chi-square | $df$ | $p$-value | Conclusion                             |
|--------------------------|-----------|------------|------|-----------|----------------------------------------|
| D(\ln PP)               |           | 6.1688     | 1    | 0.013     | Potato price is Granger cause of output supply |
| D(\ln PF)               |           | 14.963     | 1    | 0.000     | Fertilizer price is Granger cause of output supply |

| Dependent variable D(LPP) | Excluded  | Chi-square | $df$ | $p$-value | Conclusion                             |
|---------------------------|-----------|------------|------|-----------|----------------------------------------|
| D(\ln Q)                 |           | 6.980      | 1    | 0.008     | Output supply is Granger cause of potato price |
| D(\ln PF)                |           | 4.991      | 1    | 0.025     | Fertilizer price is Granger cause of output supply |

| Dependent variable D(LPF) | Excluded  | Chi-square | $df$ | $p$-value | Conclusion                             |
|---------------------------|-----------|------------|------|-----------|----------------------------------------|
| D(\ln Q)                 |           | 0.621      | 1    | 0.430     | Output supply is not Granger cause of fertilizer price |
| D(\ln PP)                |           | 0.001      | 1    | 0.968     | Potato price is not Granger cause of fertilizer price |

Source: Elaborated by the authors

### Diagnostic tests

To examine the consistency of the VEC model, residual diagnostic tests were performed. As shown in table 6, the assumptions of no serial correlation, homосedasticity, and normality were not violated.

### Table 6. VECM residual diagnostic tests

| Test                                  | Statistical value | $p$-value |
|---------------------------------------|-------------------|-----------|
| Serial correlation LM (one lag)       | $X^2 = 8.479$     | 0.486     |
| Heteroscedasticity (Joint)            | $X^2 = 47.200$    | 0.505     |
| Normality test (Joint)                | $JB = 2.713$      | 0.843     |

Source: Elaborated by the authors

In addition, as calculated by the autoregressive (AR) root diagram method, the roots of two observations were equal to one, and all others were located within the circle; therefore, the VECM satisfies the stability criteria (figure 1).
Discussion

The potato farming sector contributes significantly to the economic growth of various low- and middle-income countries. It is an essential source of income for large and small farmers in these countries, and the price factor plays a vital role in investment and planting decisions; it is thus necessary to understand its magnitude and characteristics (Soontaranurak & Dawson, 2015). However, few studies refer to the estimation of potato supply response using a VEC framework. Nonetheless, some studies have found that production is strongly motivated by price (both own-price and fertilizer price), and to a lesser extent, by non-price factors such as acreage, yield, and climate (Dlamini, 2018; Huq & Arshad, 2010).

Consequently, the results of the short- and long-run coefficient estimates obtained in the present study are considered to be significant empirical evidence that supports the existence of a relationship between potato production and both own price and fertilizer price. These findings are consistent with those of Huq and Arshad (2010); by using the VECM approach, these authors also obtained evidence for a positive relationship between potato supply and lagged own price and similar short- (0.61) and long-run (0.44) elasticity values to those found in the present study, therefore implying that price is an essential factor in farmers’ production decisions (Dlamini, 2018). In contrast with these results, Mose et al (2007), in their study on the aggregate supply response of maize to price incentives, estimated a higher short-run fertilizer price coefficient (-1.05), which would imply a more significant response to price changes.

Moreover, the estimated speed of adjustment towards long-run equilibrium found in the present study was like Huq and Arshad’s (2010). Both studies present a negative and slow adjustment rate of 0.7 and 1.18, respectively, most probably because farmers may not have access to the financial resources and technology necessary to increase production in the immediate term, thus deterring their response to incentives (Dlamini, 2018; Mose et al., 2007; Tripathi & Prasad, 2009).

Figure 1. VECM residual stability test
Source: Elaborated by the authors
Conclusions

The results of this research allow us to conclude that potato output supply is highly responsive to price incentives, which can be noted by the correct signs and the statistical significance of the short- and long-run elasticities of potato and fertilizer prices. It can be said that these positive results support the importance of the role that potato production plays in farmers’ incomes. Likewise, the Granger causality test results show a solid reciprocal causal relationship among the variables. Moreover, the error correction term indicates that, eventually, the system will converge towards equilibrium at a 70% rate, which implies that it takes a relatively long period for farmers to respond to price movements.

Therefore, the above analysis reveals that overall, the use of the VECM can contribute to improve the understanding of the long- and short-term dynamics of potato output response and thus help evaluate the effectiveness of price policies and farmers’ output decisions. Nonetheless, the proposed model could be enhanced: other lines of further research should be pursued to obtain a more accurate simulation by, for example, including factors such as expenditure in infrastructure and research and development.

Disclaimer

The authors agree with the publication of this article and declare no conflicts of interest.

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