The Impact of Remittances on the Import Demand Function in Jordan: An ARDL Bounds Testing Approach

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Abstract
The present study investigates the short- and long-run relationships between Jordan’s aggregate import demand function and its macroeconomic determinants, in addition to remittances. The study employs the autoregressive distributed lagged (ARDL) model to estimate the import function over the period 1975–2016. The preliminary statistical tests, the ADF test, confirmed that none of the variables is integrated of order 2, while the bounds testing provided evidence of the existence of a long-run equilibrium relationship between the included variables. Moreover, the diagnostic tests showed that the estimated model is free of the statistical problems. The long-run results indicated that remittances, inflation rate, and investment have a direct relationship with imports, whereas the import price index and FDI have a negative relationship. Based on these results, the study suggests that policymakers implement inflation reduction policies, increase the level of economic activities, and promote remittances inflows since they are mostly directed to investment.

Keywords: Remittances, imports, ARDL, Jordan, cointegration

Introduction
According to the geographical distribution of Jordanian imports, it is clear that China, Saudi Arabia, the United States of America, Germany, the United Arab Emirates, and Italy are the major market sources of imports, which constitute of about 46.6 percent of total imports in 2016 and around 46.7 percent in 2015 (CBJ, Annual Report, 2017, PP: 70). In terms of the share of imports by commodity in 2016, the share of crude materials and the intermediate goods make up 47.8% of total imports, while the share of consumer goods was about 34.6% and the share of capital was 16%. Hence, as these imports are very vital for economic growth, it is crucial to analyze the determinants of the import demand function in Jordan; moreover, the analysis
is crucial for policymakers in many areas, especially with regards to trade deficit (Yi-Hsien, 2012).

The objective of this study is to estimate the import demand function for Jordan by using the most modern estimation methods as well as recent data. Analyzing the import demand function is vital to any country, especially in terms of trade balance status. Among many factors affecting imports is the flow of workers' remittances to their home countries. The concern with this factor stems from its impact on the consumption of durable and non-durable imported goods. This effect is reflected in the trade balance and later in the balance of payments of the home country. Yet, to the author’s best knowledge, few existing research studies have examined the role of remittances in determining the import demand function in Jordan. For example, applying the ARDL and the bounds testing approach to cointegration over the period 1980–2015, Mugableh (2017) found that income has positive and significant impacts, while relative prices exert negative impacts on Jordan’s imports; moreover, the long-run elasticities are greater than unity. Ziad (2014) estimated the price and income elasticities of the import demand function for Jordan over the period 1980–2012 by employing the Johansen cointegration approach. He found that income and prices elasticities of imports are greater than unity. Adel and Othman (2013) estimated the import demand function for Jordan over the period 1976–2008 using multiple linear regression models. The finding indicated a direct relation between imports and GDP, CPI, and REM, whereas it is negative with relative prices and exchange rate and, in addition, they are all inelastic. Al-Hazaimeh et al. (2011) found that GDP, investment, and exports are major determinants of the import demand function for Jordan. They employed the multiple regression method for the period 1976–2008. Kreishan (2007) estimated the import demand function for Jordan over the period 1972–2004 employing the Johansen cointegration approach. The finding indicated that the aggregate import volume is price and income inelastic. As for remittances, the results show they have a positive significant impact on aggregate imports and act as a source for financing imports. Majeed (2007) estimated the traditional import demand function for Jordan over the period 1980–2004 using the dynamic OLS method. His findings revealed that relative price and income elasticities were -0.55 and 0.84, respectively. Only Adel and Othman (2013) and Kreishan (2007) investigated the role of remittances in the import function.

The current paper contributes the following. First, it analyzes the influence of on Jordan’s import demand function. Second, it uses one of the most recent modern estimation techniques, the ARDL approach, which avoids the problem of spurious regression, statistical problems, and estimation problems. Third, it utilizes up-to-date and longer time series data.
The rest of the paper is organized as follows. Section 2 presents a review of the literature on import demand function. Section 3 illustrates the econometric model specification and data. Section 4 analyzes the estimation results. Finally, Section 5 is the conclusion and suggests some policy implementation remarks.

2. Literature Review

The import demand function has been estimated by numerous applied research studies for both developed and developing countries. A large number of economic and non-economic variables were included in the model specification of the determinants of import demand function. The majority of the applied research estimates the traditional import demand function using economic activity, relative prices, real effective exchange rate, final consumption, FDI, foreign reserves, exports, and financial development among other macroeconomic variables. For example, Abdulsalam (2015) for Libya; Aldakhil and Nourah (2002) for Saudi Arabia, Al-Khulaifi (2013) for Qatar, N’guessan and Yue (2010) for Cote D’Ivoire, Khurram and Syed (2012) for Pakistan, Nazif and Jaehyuk (2015) for Turkey, Emmanuel and Mooya (2013) for Namibia, Ibrahim and Ahmed (2017) for Sudan, BigBen (2016) for Nigeria, Zhou and Dube (2011) for CIBs countries, Sulaiman and Saba (2016), and AbdulRashid and Tayyaba (2010) for Pakistan. All the above-mentioned studies use different estimations methods indicating a positive association between imports and income, and a negative association with relative prices and real effective exchange rate.

Following the scope of the current study, this section reviews the most recent studies to select the appropriate and relevant factors to estimate the import demand function for Jordan, whereas the emphasis is on the impact of remittances on imports.

Chantha et al. (2018) estimated the long-run and short-run import demand function for Cambodia over the period 1993–2015 by employing the ARDL model. Their empirical finding showed that inflation and the exchange rate have negative impacts, whereas exports have a positive impact. Using the standard OLS regression approach over the period 1988–2015 for Saudi Arabia, Abdullah Almounser (2017) found that GDP, government expenditures, private consumption, and investment have positive and significant impacts, while the real effective exchange rate exerts a negative impact. Applying the ARDL model to data over the period 1973–2013 for Pakistan, Sulaiman and Saba (2016) found that consumption, exports, and investment have a positive significant impact; final consumption expenditure, and government consumption expenditure showed negative and significant impact. Ahmed et al. (2014) examined the short-run and long-run relationships between imported goods and workers’ remittances in Pakistan over the period
2008–2012 employing Johansen cointegration and Granger causality. The finding showed a positive and significant impact of remittances on imports. However, Granger causality indicated a unidirectional causality runs from imports to remittances. M. Sayed (2014), applying the vector error correction (VECM) model to annual data over the period 1991–2011 for Egypt, found that remittances exert a positive and significant impact on imports as indicated by the unidirectional causality runs from remittances to imports. Dewan et al. (2013) applied the Johansen cointegration approach to monthly data over the period 2005–2011 for Bangladesh. The finding showed that remittances have an insignificant impact on imported goods, and a unidirectional causality runs from imports to remittances. Guna (2013), applying cointegration and a VECM model to monthly data over the period 2001–2011 for Nepal, found that remittances exert a significant positive impact on imported merchandised goods and services, where the unidirectional causality runs from remittances to imports. Karan and Sanjanya (2013), employing the OLS method and Granger causality test for Nepal over the period 2001–2009, found that remittances Granger-cause imports. Soana and Olta (2013) adapted a VECM model using monthly data over the period 1999–2011 for Albania. The finding showed that GDP and remittances exert positive impacts on imports, while real effective exchange rate and average tariffs have negative impacts. Yi-Hsien (2012) applied the ARDL approach for data over the period 1992–2011 for China. The finding indicated that GDP has a significant positive impact, whereas real effective exchange rate was negative and insignificant. Using the OLS method, Munir et al. (2007) estimated the import function for Pakistan. The finding showed that remittances and GDP have positive and significant impacts on imports, while real effective exchange rate was negative. Khair and Nazakat (2005) applied the OLS method using quarterly data over the period 1975–2004 in Pakistan. The findings indicated that remittances have a positive and significant impact on imports.

The surveyed literature pointed out the positive impact of remittances on aggregate imports and that the impact size differs among receiving countries depending on the estimation methods and data span. Table (1) summarizes the findings of the surveyed literature.
### Table (1): Most Frequent Used Variables in the Literature Survey

<table>
<thead>
<tr>
<th>Variable</th>
<th>Positive</th>
<th>Negative</th>
<th>Insignificant</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>INFLATION</strong></td>
<td>Aldakhil and Nourah (2002)</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
3. Methodology: Data and Model Specification

This section addresses the econometric methodology and the data sources adapted in the current study to estimate the import demand function in Jordan. Moreover, it illustrates the model specification in terms of the variables to be included in the model.

3.1 Model Specification

Chanth a et al (2018) indicated that the traditional import demand function is based on the imperfect substitution theory, which focused on the role of the importing country’s income, the price the imported goods, and the import substitute goods. The current study utilizes the previous reviewed literature which shows different forms of the import demand function based on single-country, groups-countries, and various econometric approaches to estimate the import demand function to choose the relevant variables relevant to Jordan’s economy case. Accordingly, the functional form of the import demand function in Jordan is as follows:

\[ IM = f(Y, CPI, IPI, REM, FDI, GFCF) \]  \hspace{1cm} (1)

Where IM is the import demand; Y is the real gross domestic product measured at constant prices ($2005=100$); REM is the volume of formal remittances; CPI is the consumer price index (2005=100) proxy for the inflation rate; IPI is the import price index; GFCF is gross fixed capital formation proxy for investment; and FDI is the stock of foreign direct investment. It is expected that the real GDP to exert positive impact on imports, since the increase in the real GDP stimulates private consumption of imports. As for the rate of inflation, the model expects a positive impact on imports; as domestic inflation rate increase, people shift to imports which are cheaper. Remittances are expected to have positive impact on imports. Remittances can be used either for consumption or investment activities, which increases the demand for goods including imported ones. The import price index is expected to have negative impact; while GFCF is expected to exert positive impact. FDI can have negative or positive impact on imports. The long-run import demand function for Jordan in is expressed in logarithmic form.

3.2 Data Description

The required data for the estimation process were obtained from various sources, the UNTCAD, the Central Bank of Jordan publications, and the World Development Indicators (WDI).

3.3 Econometric Analysis: ARDL bounds testing

The objective of the current study is to estimate the long-run and short-run relationships between the variables of the import demand function for

The study employs one of the most widely used econometric methods in time-series analysis, the autoregressive distributed lagged (ARDL) model bounds testing approach to cointegration introduced by Pesaran, et al. (2001) to estimate the long-run and short-run relationships between Jordan's aggregate import and a set of explanatory variables. The analysis involves examining the degree of integration of the series via the unit root test, the cointegration test to examine the existence of long-run equilibrium relationships, and the Granger causality test within a VECM framework. There are many advantages for adapting such approaches over other procedures proposed by Johansen (1990, 1991) and Engle (1987). First, it is suitable irrespective of the order of integration of the variables; either I(0) or I(1), as long as it is not I(2); therefore, this would avoid the stationarity problems (Zhou and Dube, 2011). Second, in the case of using small samples, the approach is more appropriate than other cointegration approaches.

Generally, the first step is to test for the stationarity properties of all variables before proceeding with the ARDL bounds testing to ensure that all time series are either I(0) or I(1) but not I(2). Therefore, the ARDL bounds testing approach is employed to estimate equation (2) using OLS to test for the presence of a long-run equilibrium relationship among the variables

\[ \Delta IM = \alpha_0 + \sum_{i=1}^{n} \alpha_1 \Delta IM_{t-1} \]

\[ + \sum_{j=0}^{s} \beta_2 \Delta Y_{t-i} \]

\[ + \sum_{j=0}^{p} \beta_3 \Delta CPI_{t-j} \]

\[ + \sum_{j=0}^{p} \beta_4 \Delta REM_{t-j} + \sum_{j=0}^{p} \beta_5 \Delta IPI_{t-j} + \sum_{j=0}^{p} \beta_6 \Delta GFCF_{t-j} \]

\[ + \sum_{j=0}^{p} \beta_7 \Delta FDI_{t-j} + \delta_1 IM_{t-1} + \delta_2 Y_{t-1} + \delta_3 CPI_{t-1} \]

\[ + \delta_4 REM_{t-1} + \delta_5 IPI_{t-1} + \delta_6 GFCF_{t-1} + \delta_7 FDI_{t-1} + \varepsilon_t \quad (2) \]
Where $\Delta$ is the first difference operator; $\alpha_0$ is the intercept; $p$ is the maximum lag length; $i$ is the number of lags; $\beta(i, p = 1, \ldots, 7)$ indicates the short run coefficients; $\delta(i, p = 1, \ldots, 7)$ shows the long-run coefficients; and $\epsilon_t$ is the white noise error term.

The hypothesis to test the presence of long-run relationship among the model variables is set as following:

$$H_0: \delta_1 = \delta_2 = \delta_3 = \ldots = \delta_7 = 0$$
$$H_1: \delta_1 \neq \delta_2 \neq \delta_3 \neq \ldots \neq \delta_7 \neq 0.$$ 

The null hypothesis was tested by performing an F-test for the joint significance of the coefficients of the lagged levels of the variables against the critical values introduced by Narayan (2005). If the F-statistic is greater than the upper bound critical value, then the null hypothesis of no cointegration can be rejected. Conversely, if the F-statistic is less than the lower bound critical value, then we cannot reject the null hypothesis of no cointegration. However, when the f-statistic is within the two bounds, then the test is inconclusive. The optimal lag-length of the ARDL model is selected using the Akaike information criteria (AIC).

If there is evidence on the existence of long-run relationship between import demand and its determinants, the next step is to examine the short-run dynamic coefficients and the $ECT_{t-1}$ coefficient. The $ECT_{t-1}$ coefficient measures the speed of adjustment from short-run towards long-run equilibrium among variables (Chantha et al., 2018). Then, the short-run relationship for Jordan’s import demand can be expressed as following in equation (3):

$$\Delta IM = \alpha_0 + \sum_{i=1}^{n} \alpha_{1i} \Delta IM_{t-1} + \sum_{j=0}^{p} \alpha_{2i} \Delta Y_{t-j}$$
$$+ \sum_{j=0}^{p} \alpha_{3i} \Delta CPI_{t-i}$$
$$+ \sum_{j=0}^{p} \alpha_{4i} \Delta IPI_{t-j}$$
$$+ \sum_{j=0}^{p} \alpha_{5i} \Delta REM_{t-j} + \sum_{j=0}^{p} \alpha_{6i} \Delta FDI_{t-j}$$
$$+ \sum_{j=0}^{p} \alpha_{7i} \Delta GFCF_{t-j} + \lambda_1 ECT_{t-1} + \epsilon_t \ (3)$$

Where $\Delta$ is the first difference operator; $\alpha_0$ is the constant; $p$ is the maximum lag length; $i$ is the number of lags; $\alpha(j, i = 1, \ldots, 7)$ indicates the
short run coefficients; and $\lambda$ is the coefficient of the lagged error term, $ECT_t\equiv\lambda$, and it should be negative.

3.4 Diagnostic stability tests

Diagnostic statistics are adapted to ensure the validity of the estimation results. For that reason, the Lagrange multiplier (LM) test of residual serial correlation, Ramsey's RESET test for functional form misspecification, normality test, and White's test for heteroscedasticity are employed. To test for structural changes, the stability of the estimated short-run and long-run coefficients was examined by employing the cumulative sum of recursive residuals (CUSUM) and cumulative sum of squares of recursive residuals (CUSUMSQ) tests proposed by Brown et al. (1975).

4. Estimation results and discussion

4.1. Unit root test (Stationarity test)

According to Pesaran et al. (2001), the first step before proceeding with the ARDL bounds testing is the determination of the order of integration to ensure that the time series are either I(0) or I(1) but not I(2). Table (2) reports the results of the augmented Dickey-Fuller (1979) unit roots test. The ADF test results indicate that all variables are stationary at their first-differenced integrated of order one, I(1).

<table>
<thead>
<tr>
<th>Variable</th>
<th>ADF results (level)</th>
<th>ADF results (differenced)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Constant</td>
<td>Constant &amp; Trend</td>
</tr>
<tr>
<td>LIM</td>
<td>-0.3 (0)</td>
<td>-2.28(1)</td>
</tr>
<tr>
<td>IP</td>
<td>0.14(0)</td>
<td>-2.26(3)</td>
</tr>
<tr>
<td>LFDI</td>
<td>-1.37(0)</td>
<td>-2.02(3)</td>
</tr>
<tr>
<td>LY</td>
<td>-1.49(1)</td>
<td>-1.83(1)</td>
</tr>
<tr>
<td>CPI</td>
<td>0.89(0)</td>
<td>-2.45(2)</td>
</tr>
<tr>
<td>LRMIT</td>
<td>-1.24(0)</td>
<td>-3.92(4)</td>
</tr>
<tr>
<td>GFCF</td>
<td>-3.69(0)</td>
<td>-3.92(6)</td>
</tr>
</tbody>
</table>

(*), (**), (***), significant at 1%, 5%, and 10% respectively, lags numbers are in Parenthesis

Adapting the ADF results moves the analysis to the next step, which is testing for the existence of long-run equilibrium relationship among variables.

4.2 Diagnostic and stability tests

The statistical tests of the ARDL (2, 3, 2, 0, 3, 3, 1) estimation results are necessary to ensure that the model is free of statistical problems. Diagnostic tests for serial correlation, functional form, normality, and heteroscedasticity have been conducted and the results are presented in Table (4). The LM Serial correlation test, in addition to DW test (2.16) indicates that
the model does not suffer from the problem as it is shown by the insignificant value of LM F-statistic test (1.15), therefore, one may accept the hypothesis of no serial correlation. Additionally, the Breusch-Pagan-Godfrey heteroscedasticity test of the insignificant F-statistics test (0.42) indicates the absence of this problem; and Ramsey's RESET of Functional form test is insignificant. Finally, Jaque-Bera’s normality test statistic is insignificant (0.809), revealing that error terms are normally distributed. Based on these test results, the model is free of econometric problems; and the estimation results are valid for meaningful interpretation.

<table>
<thead>
<tr>
<th>Test</th>
<th>F-statistics</th>
<th>P-Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Serial Correlation</td>
<td>2.8</td>
<td>0.108</td>
</tr>
<tr>
<td>Functional Form</td>
<td>0.66</td>
<td>0.59</td>
</tr>
<tr>
<td>Normality</td>
<td>0.70</td>
<td>0.702</td>
</tr>
<tr>
<td>heteroscedasticity</td>
<td>0.42</td>
<td>0.6</td>
</tr>
</tbody>
</table>

The study applies the cumulative sum of recursive residuals (CUSUM) and the CUSUM of the square (CUSUMSQ) to ensure the parameters’ stability. Figure 1 shows that the plots of the CUSUM and CUSUMSQ statistic fall with the critical bands of the 5 percent confidence interval of parameter stability. Therefore, the results confirm the existence of the stability in the parameters over the study period.

4.3 Cointegration results

The next step is to examine the existence of the long-run relationships between model variables by applying the bounds testing approach to cointegration. As Table (5) shows, the calculated F-statistics of 4.02 is greater than the upper bound critical value of 4.01 provided by Pesaran (2001) at the 5 percent level, and hence, one can reject the null hypothesis of no cointegration. The result from bounds testing approach to cointegration provides evidence on the long-run relationship between the variables.
<table>
<thead>
<tr>
<th>Variable</th>
<th>Coefficient</th>
<th>Std. Error</th>
<th>t-Statistic</th>
<th>Prob.</th>
</tr>
</thead>
<tbody>
<tr>
<td>DLOG(IM(-1))</td>
<td>0.412124</td>
<td>0.237354</td>
<td>1.736326</td>
<td>0.1081</td>
</tr>
<tr>
<td>DLOG(Y)</td>
<td>3.066670</td>
<td>0.677223</td>
<td>4.528300</td>
<td>0.0007</td>
</tr>
<tr>
<td>DLOG(Y(-1))</td>
<td>1.953987</td>
<td>0.913596</td>
<td>2.138786</td>
<td>0.0537</td>
</tr>
<tr>
<td>DLOG(Y(-2))</td>
<td>-1.915532</td>
<td>0.302432</td>
<td>-6.333766</td>
<td>0.0000</td>
</tr>
<tr>
<td>DLOG(REM)</td>
<td>0.227843</td>
<td>0.143270</td>
<td>1.590309</td>
<td>0.1378</td>
</tr>
<tr>
<td>DLOG(REM (-1))</td>
<td>0.295041</td>
<td>0.174395</td>
<td>1.691798</td>
<td>0.1165</td>
</tr>
<tr>
<td>D(CPI)</td>
<td>0.012202</td>
<td>0.003128</td>
<td>3.900847</td>
<td>0.0021</td>
</tr>
<tr>
<td>D(IPI)</td>
<td>0.002770</td>
<td>0.000864</td>
<td>3.205518</td>
<td>0.0076</td>
</tr>
<tr>
<td>D(IPI(-1))</td>
<td>0.000813</td>
<td>0.000900</td>
<td>0.903085</td>
<td>0.3843</td>
</tr>
<tr>
<td>D(IPI(-2))</td>
<td>0.001459</td>
<td>0.000314</td>
<td>4.650469</td>
<td>0.0006</td>
</tr>
<tr>
<td>DLOG(GFCF)</td>
<td>0.236315</td>
<td>0.159154</td>
<td>1.484821</td>
<td>0.1634</td>
</tr>
<tr>
<td>DLOG(GFCF(-1))</td>
<td>-0.434197</td>
<td>0.258588</td>
<td>-1.679104</td>
<td>0.1190</td>
</tr>
<tr>
<td>DLOG(GFCF(-2))</td>
<td>0.453068</td>
<td>0.152968</td>
<td>2.961846</td>
<td>0.0119</td>
</tr>
<tr>
<td>DLOG(FDI)</td>
<td>-0.011923</td>
<td>0.016648</td>
<td>-0.716191</td>
<td>0.4876</td>
</tr>
<tr>
<td>CointEq(-1)</td>
<td>-1.557629</td>
<td>0.226066</td>
<td>-6.890148</td>
<td>0.0000</td>
</tr>
</tbody>
</table>

The important feature of these results is the error correction term which turned out to be negative and significant at the 1% level. This result provides evidence of the presence of a long-run causality runs from explanatory variables to (IM). The coefficient of the error correction term is (−1.56) indicating that 1.56% of short-run shock is corrected in each year, and it takes less than a year to restore long-run equilibrium.
Table 7 reports the long-run estimation results of ARDL (2, 3, 2, 0, 3, 3, 1). The table shows that the economic activity (GDP) is positive and statistically significant at the 1 percent level with elasticity equal to 1.4%. The result indicates that a 1% increase in the GDP increases imports by 1.4%. The result of economic growth or economics activity is in line with Yi-Hsien (2012) for China, Zhou and Dube (2011) for CIBs, Soana and Olta (2013) for Albania, Munir et al. (2007) for Pakistan, Abdul Rashid and Tayyaba (2010) for Pakistan, Ibrahim and Ahmed (2017) for Sudan, Khurram and Syed (2012) for Pakistan, and Aldakhil and Nourah (2002) for Saudi Arabia who found a significant positive impact of GDP (income) on the import demand function. On the other hand, in the Nigerian case, BigBen (2016) found an insignificant impact of GDP. As for Jordan, Mugableh (2017), Ziad (2014), Adel and Othman (2013), and Al-Hazaimeh (2011) found a direct relation between GDP and imports. The results indicated that remittances have a significant positive impact on imports, where the elasticity of imports with respect to remittances is inelastic. The elasticity is 0.218, hence, an increase of remittances by 1% leads to an increase in imports by 0.218%. The remittances result is in line with Soana and Olta (2013) for Albania, Ahmed et al. (2014) for Pakistan, Karan and Sanjanya (2013) and Dewan et al. (2013) for Bangladesh, M. Sayed (2014) for Egypt, Gunna (2013) for Nepal, Khair and Nazakat (2005) and Munir et al. (2007) for Pakistan, Adel and Othman (2013) and Kreishan (2007) for Jordan, who all found a significant positive impact; whereas BigBen (2016) found insignificant impact of remittances on imports for Nigeria. The price level exerts a very low positive and significant impact and is inelastic as well. An increase in the price index by 1% leads to a 0.007% increase in imports. The price index result is in line with Aldakhil and Nourah (2002). The import price index has a negative but insignificant impact on imports; in addition, the influence is very weak (-0.0006). Therefore, a 1% increase in the imports price index leads to a 0.0006% decrease in imports. The import price index result is in line with Aldakhil and Nourah (2002) for Saudi Arabia. The investment level measured as the gross fixed capital formation (GFCF) has a significant positive impact on imports, and it is inelastic. A 1% increase in GFCF increases imports by a 0.49%. This result is in line with Karan and Sanjanya (2013) for Nepal, Sulaiman and Saba (2016) for Pakistan, Emmanuel and Mooya (2013) for Namibia, Nazif and Jaehyuk (2015) for Turkey, and N’guessan and Yue (2010) for Cote D’Ivoire, whose results supported the significant positive impact of investment on imports.
Finally, the foreign direct investment (FDI) has a significant negative impact on imports as well as inelastic. A 1% increase in FDI decreases imports by 0.034%. This result contradicts the insignificant impact found by Chanthan et al. (2018) for Cambodia and Sulaiman and Saba (2016) for Pakistan.

5. Conclusion and Policy Implication

The present study is an attempt to examine the impact of remittances along with other macroeconomic variables on the imports of Jordan over the period 1976–2016. It adapted the most recent estimation technique, the autoregressive distributed lagged (ARDL) model, which has many advantages over other techniques. The estimation results support the existence of the long-run equilibrium relationship between them. The ECM coefficient is negative and significant indicating that the causality runs from the explanatory variables to imports and, in addition, the magnitude of the coefficient (1.56) shows that it takes less than one year to restore the long-run equilibrium after a short-run shock. The long-run results show a significant positive influence of remittances, level of income (economic growth), investment, and price level on imports; meanwhile, the import price index and FDI have negative and significant impacts on imports. In addition, all variables are inelastic except remittances. This pointed out the important role of remittances on stimulating imports, and the ultimate effect depends on the type of imported goods—durable (capital) or non-durable (consumption). The current study suggests some policy recommendations that are expected to help policymakers adapt some policy measures to reduce the trade balance consequences. First, reducing the price level would help reduce imports and increase exports, and hence, improve trade balance. Second, increasing the level of economic activities and economic growth would increase capital goods for investment.

The literature on the import demand function contains a considerable number of macroeconomic variables that are expected to influence the import demand function. The present study utilized selected macroeconomic variables that are thought to be crucial to import demand function. Therefore, the study recommends that future research on the import demand function by
investigating new variables, different estimation method, and data to compare the present results

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33. Soana, Jaupillari (Teka), and Olta Zoto (2013). *Assessment of Demand for Imports through the VECM Model; Journal of Knowledge Management, Economics and Information Technology*, 3(6), 1-17.


