

Modelling the Symmetrical and Asymmetrical Effects of Global Oil Prices on Local Food Prices: A MENA Region Application

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1 **Modelling the symmetrical and asymmetrical effects of global oil** 2 **prices on local food prices: A MENA region application**

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4 **Abstract:** This paper explores the complex nexus between the global oil prices and the food
5 prices of Middle East and North Africa (MENA) region during the period 2000–2020. Both
6 linear and nonlinear models of the autoregressive distributed lag (ARDL) approach are adapted
7 into panel data form to investigate the symmetrical and asymmetrical influence of oil prices on
8 food prices. The key results are summarized: i) The effect of oil prices on food prices is
9 significantly positive including both oil-exporting and oil-importing nations are verified in the
10 long-term. The positive impact on oil-exporters—due to higher oil revenues—is greater than
11 importing nations, leading to an increased demand for food. Additionally, the effect on oil-
12 exporters is negative and significant in the short-term but not significant for importers. ii) The
13 panel analysis for the MENA sample confirms the presence of negative short-term asymmetric
14 behaviour, while in the long-term, the asymmetric effect is positive, indicating that food prices
15 increase regardless of fluctuations in oil prices. iii) Wald test results support asymmetric co-
16 integration for the whole sample of the MENA due to the heterogeneous response within the
17 oil-importing and exporting samples. Specifically, the non-linear ARDL test results affirm the
18 absence of an asymmetric nexus among oil and food prices for oil-exporting group (including
19 Saudi Arabia, Saudi Arabia, United Arab Emirates) and Tunisia within the oil-importing group.
20 Although there are differences in the direction and degree, the food prices of other countries

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21 are asymmetric to the oil price. This study provides recommendations that are useful to MENA
22 countries to establish a stable mechanism for oil and food prices to ensure food security in the
23 region.

24 **Keywords:** Oil prices; Food prices; MENA region; Panel ARDL; Symmetry and Asymmetry.

25 **Introduction**

26 As important strategic resources, food and oil have been widely studied by scholars (Dalheimer
27 et al., 2021; Mokni and Ben-Salha, 2020; Sarwar et al., 2020). In recent years, international
28 food prices have continued to rise. Many reasons contribute to rising prices of food; however,
29 high oil prices are a major factor. The expansion of biomass energy—such as biodiesel and
30 bioethanol—in response to climate change has caused a crowding-out effect on food production
31 for people, while high oil prices have increased agricultural costs incurred in the production
32 and transportation processes (Gardebroek and Hernandez, 2013; Sarwar et al., 2020; Dalheimer
33 et al., 2021). Rising food prices have posed a substantial threat to food security especially for
34 countries that rely heavily on importing food. The UN Secretary-General Ban Ki-Moon noted
35 that global food prices rose by more than 50% in 2007 alone. This severe situation, and the
36 double crisis caused by rising energy prices, threatened the international community's plan to
37 achieve their Millennium Development Goals in 2015. It is estimated that if food prices increase
38 50–80%, many people will face malnutrition and hunger. Consequently, recognizing the
39 internal relationship among food oil and oil prices is an important prerequisite to stabilizing
40 food prices and establishing an early warning mechanism. However, due to the huge differences
41 in terms of resource endowment, agricultural productivity, and economic status among regions
42 of the world, the relationship between the two prices can have distinguish mechanism. For
43 example, countries rich in agricultural products will be much less affected by high oil prices
44 than countries with poor supply capacity of agricultural products. Therefore, the internal
45 relationship between food and oil prices should be analysed according to the research objective.

46 As for the region of Middle East and North Africa (MENA), researching how oil prices affects
47 food prices is even more important. The main reasons are: 1) Most countries are net importers
48 of food commodities as food crops are difficult to cultivate due to insufficient water supply and
49 limited access to arable land. They are particularly vulnerable to fluctuations in international
50 commodity markets and are most seriously affected by rising prices. Therefore, stable food
51 prices and reliable sourcing of imports are the basis for food security in these countries; 2) Some
52 countries that are rich in oil resources but lacking in food have implemented the strategy of the
53 ‘Oil for Food programme’. The rising oil price will attract great economic benefits, but more
54 will have to be spent on buying food. Thus, the social welfare of these countries is uncertain
55 given the volatility of oil and food prices; 3) Although MENA countries are geographically
56 concentrated, the resources and economic conditions of these countries are vastly different.
57 Only certain countries are rich in oil products. Therefore, for oil-importing countries (*oil-*
58 *importers*) and oil-exporting countries (*oil-exporters*), international oil prices can exhibit
59 differing impact mechanisms on local food prices. Researching the influence of oil prices on
60 the MENA region's food prices—where food is primarily dependent on imports—is of great
61 significance in establishing a stable food price mechanism and to ensure regional security.
62 Consequently, this study will focus on exploring the distinctive mechanisms impacting oil and
63 food prices in the MENA region by comparing *oil-exporters* to *oil-importers*. This research is
64 among the first to focus on examining this from a MENA perspective. Previous studies often
65 using time series which cannot capture inter-group information and eliminate the individual
66 fixed effect; this study explores such relationship using a panel data sample. Considering food
67 prices may appear similar or dissimilar as oil prices fluctuate, the research builds linear and
68 non-linear Autoregressive distributed lag (ARDL) modelling method to address symmetrical and
69 asymmetrical inter linkage between the price food and oil. The contributions of this study are:
70 i) focusing on a specific group of countries in the MENA region to determine the impacts of

71 the oil price on their food prices; ii) Focusing on a comparative exploration on the subject by
72 comparing oil exporters with oil importers. iii) Simultaneously, the symmetrical and
73 asymmetrical interactions in the food-oil price nexus are captured by applying panel data from
74 oil-exporters and importers of the MENA region. The remainder of this paper has the following
75 structures: in section 2 the research literature relating to food-oil prices nexus is summarized;
76 Section 3 highlights the panel symmetric and asymmetric ARDL models; in section 4 the data
77 and indicators used for the study are presented; Section 5 discusses the empirical application
78 for MENA countries from 2000 to 2020; and the paper concludes with Section 6, which
79 discusses policy implications.

80 **Literature review**

81 Substantial research on food and oil markets was stimulated by the 2008 global food crisis. The
82 volatility of the crude oil price affects food commodities through increasing fertilizer costs,
83 transportation costs, and fuels for agricultural machinery and so on (Chen et al. (2019). The
84 existing studies contends that food prices are significantly and positively related to the oil price.
85 For instance, Chen et al. (2010) using the ARDL model, detected that crude oil fluctuations and
86 other grain market fluctuations have a significant impact on the price of grain. Esmaeili and
87 Shokoohi (2011) using a principal component analysis to investigate the nexus between the
88 global oil price, world prices of food and other macroeconomic variables. Their findings
89 revealed that the price of oil influences food prices in an indirect way, as confirmed by Pal and
90 Mitra (2018), Gohin and Chantret (2010), and Ciaian and Kancs (2011). However, some
91 scholars have found that the food prices is not influenced by oil price fluctuations. For example,
92 Reboredo (2012) examined co-movements among the global oil price and those of wheat, corn,
93 soybeans by applying copulas. His results indicated that the increase in food price was not
94 caused by volatility in oil prices. Therefore it supports neutrality against oil price fluctuations
95 in agricultural markets. Baumeister and Kilian (2014) confirmed that shocks in oil prices are

96 not linked to food prices in the United States due to the small contribution of agricultural
97 commodities to total food prices. Another analysis containing the volatility transmission of
98 corn, oil and ethanol prices was performed by Gardebroek and Hernandez (2013). They found
99 there is no major fluctuations in the energy market that boost price variability on the corn
100 market.

101 Due to the differences in the selected samples and the complexity of the relationships between
102 them, it is difficult to reach a consensus from the literature. Increasingly, scholars believe this
103 is a complex nonlinearity relationship. Using Malaysia as a case, Ibrahim (2015) reported the
104 presence of short- and long-run asymmetries in food prices behaviour when oil price increases
105 led to increasing food prices— but there was no association between the decline in oil prices
106 and food prices. The potential asymmetry for transit from the price of oil to food is also analysed
107 in an Indian context (Pal and Mitra, 2016). For an oil-dependent emerging economy as Nigeria,
108 Nwoko et al. (2016) investigated the causal link between food and oil prices volatility and
109 concluded that the causality from oil prices to aggregate food price fluctuations is
110 unidirectional. Olayungbo and Hassan (2016) examined the symmetric interactions between
111 food and oil prices between oil and food prices of *oil-exporters* by applying a panel ARDL
112 approach. The long-run finding indicated that oil prices affected food prices positively and
113 significantly while the short-term impact was similarly positive and significant. These results
114 were verified by Meyer et al. (2018).

115 As the hub connecting Asia, Europe, and Africa and the most important oil-producing region
116 in the world, the MENA region has garnered much attention. Ek Fälth et al. (2020) explored
117 the impact of nuclear energy, land availability, and the expansion of transmission on the cost
118 of electricity from renewable sources by comparing MENA region to Europe. Kassouri and
119 Altıntaş (2020) examined the trade-off regarding environmental and human well-being issues.
120 Bellakhal et al. (2019) focused on the interaction impact between governance and trade

121 openness on investments in renewable energy in the MENA region over the period 1996-2013.
122 Apergis et al. (2014) investigated the Dutch disease impact of oil rents on added value for the
123 agricultural sector in oil-producers of MENA region, establishing a long-run negative
124 relationship that relatively slowly re-equilibrium with an added agricultural value following a
125 boom in oil rents. Although previous literature has investigated many aspects of the MENA
126 region, the mechanism influencing food and oil prices has been neglected.

127 In conclusion, although there are considerable researches on the effects of oil price on food
128 price, there are few studies that examine and explore how oil price influences food prices using
129 a sample of MENA nations. Previous studies on this issue have always used time series which
130 cannot capture group information or mitigate individual fixed effects. Consequently, this study
131 explores this relationship using panel linear and non-linear ARDL models to assess both short-
132 term and long-term dynamic interactions between food and oil prices for MENA countries.

133 **Indicators and data**

134 To explore the association of food and oil prices in MENA countries, this study employs panel
135 data of food prices (*FDP*) from 2000 to 2020 for 14 countries of the MENA region and annual
136 time series data of the global oil price (*OLP*). Referring to previous literature (e.g. Meyer et al.,
137 2018; Olayungbo and Hassan, 2016; Taghizadeh-hesary et al., 2019), additional panel data such
138 as the inflation rate (*INF*), the degree of trade openness (*TO*), and urbanization levels (*URB*)
139 are chosen as control variables. Given data availability, we only obtain time series annual data
140 for the global oil price (the average of two major types of crude oil: Brent and WTI). Food
141 prices in the MENA region are transformed by consumer prices indices to actual values for the
142 base period (2015 = 100). The degree of trade openness (*TO*) is expressed as a percentage of
143 total imports and exports to Gross Domestic Product (GDP); and the urbanization level (*URB*)
144 is measured as the urban population ratio. The relevant data was derived from the Energy
145 Information Administration (EIA), the (BP) Statistical Review of World Energy, the Food and

146 Agriculture Organization (FAO), the World Bank database. Specific indicators for variables are
 147 shown in Table 1.

148 **Table 1** Specific indicators for variables

Dimension	Indicator	Symbol	Log form symbol	Data source
Local Food prices	Food consumer prices	FDP	LFDP	FAO
Global oil price	The Brent and WTI average	OLP	LOLP	BP,EIA
The degree of trade openness	The percentage of total imports and exports to GDP	TO	LTO	The World Bank Database
Inflation rate	The percentage rate of change in prices level	INF	LINF	The World Bank Database
Urbanization level	The ratio of urban population to the total population	URB	LURB	The World Bank Database

149 The sample countries were divided into two groups to analyse the data more precisely:

150 Group 1: Oil-exporters group (Algeria, Bahrain, Iraq, Kuwait, Oman, Qatar, Saudi Arabia and
 151 UAE). Group 2: Oil-importers group (Jordan, Lebanon, Tunisia, Morocco, Egypt, West Bank and
 152 Gaza).

153 Table 2 provides descriptive statistics. For the *oil-exporters* sample, the mean of food prices
 154 (*FDP*) is greater than for *oil-importers* (81.702). The oil price (*OLP*) has a volatility of 25.704
 155 based on its standard deviation, which denotes some shocks between 2000 and 2020. Inflation
 156 rate (*INF*) among the oil-exporting group has a maximum value of 53.231% and minimum of -
 157 10.067%, whereas those values for oil-importers reach 29.602% and -3.749%, respectively. The
 158 degree of trade openness (*TO*) among the MENA region has a mean of 92.768 %, with a
 159 maximum of 191.872%, and a minimum of 30.247%. Finally, the mean of the urbanization level
 160 (*URB*) for *oil-exporters* (83.294%) is higher than for *oil-importers* (69.049%). Overall, different
 161 connected mechanisms for food and oil prices in the two types of nations are clearly indicated.

162 **Table 2** Statistics of descriptive variables

Variable	Obs.	Min.	Max.	Mean	Std.Dev.
Oil- exporting countries					
<i>FDP</i>	168	33.342	122.646	83.885	19.138
<i>OLP</i>	168	25.325	103.27	63.161	25.704
<i>INF</i>	168	-10.067	<u>53.231</u>	3.616	6.596

<i>TO</i>	168	50.045	191.872	100.244	32.154
<i>URB</i>	168	59.919	100	83.294	11.606
Oil-importing countries					
<i>FDP</i>	126	27.717	193.637	81.702	28.467
<i>OLP</i>	126	25.325	103.27	63.161	25.73
<i>INF</i>	126	-3.749	29.602	4.414	5.149
<i>TO</i>	126	30.247	144.881	82.053	24.189
<i>URB</i>	126	42.701	92.203	69.049	15.755
MENA countries					
<i>FDP</i>	294	27.717	193.637	82.95	23.572
<i>OLP</i>	294	25.325	103.27	63.161	25.671
<i>INF</i>	294	-10.067	53.231	3.958	6.022
<i>TO</i>	294	30.247	191.872	92.448	30.335
<i>URB</i>	294	42.701	100	77.189	15.249

163 Table 3 shows the correlation matrix among variables. For the MENA countries there is a
164 positive correlation between food prices (FDP) and oil prices (OLP). The inflation rate (INF)
165 is positively correlated with food prices (FDP), except for *oil-exporters* where there is a
166 negative correlation. Furthermore, the degree of trade openness (TO) negatively correlated with
167 food prices (FDP) for oil-importers and the MENA region, and the relation between
168 urbanization level (URB) and food prices (FDP) is positive for *oil-exporters*. These results
169 indicate there is no potential multi-collinearity problem.

170 **Table 3** Matrix of correlations among variables

Variable	<i>FDP</i>	<i>OLP</i>	<i>INF</i>	<i>TO</i>	<i>URB</i>
Oil- exporting countries					
<i>FDP</i>	1.000				
<i>OLP</i>	0.427	1.000			
<i>INF</i>	-0.270	0.063	1.000		
<i>TO</i>	0.001	0.167	0.053	1.000	
<i>URB</i>	0.183	0.062	-0.200	0.298	1.000
Oil-importing countries					
<i>FDP</i>	1.000				
<i>OLP</i>	0.286	1.000			
<i>INF</i>	0.331	0.172	1.000		
<i>TO</i>	-0.046	0.224	-0.297	1.000	
<i>URB</i>	0.128	0.061	-0.458	0.599	1.000
MENA countries					
<i>FDP</i>	1.000				
<i>OLP</i>	0.346	1.000			
<i>INF</i>	0.006	0.102	1.000		
<i>TO</i>	-0.005	0.177	-0.071	1.000	
<i>URB</i>	0.154	0.054	-0.298	0.486	1.000

171 Methodology

172 This article specified the recent method proposed by Pesaran et al. (2001) of the autoregressive
173 distributed lag (ARDL) boundary testing to determine the influence of global oil prices on
174 MENA countries' local food prices. This approach has many advantages compared to classical
175 co-integration models (e.g. Engle and Granger, 1987; Johansen and Juselius, 1990), including:
176 i) Estimations of both short- and long-term coefficients can be captured simultaneously; ii) it is
177 practicable even if I(0) or I(1) or combination of any of the regressors are used ; iii) prevent
178 endogeneity issues by taking into consideration a small sample and producing better outcomes
179 over other co-integration methods. All variables are logarithmically transformed to address the
180 potential heteroskedasticity problem. We construct panel linear ARDL and non-linear ARDL
181 models according to Shin et al. (2014) and Salisu and Isah (2017) to detect the existence of
182 symmetrical and asymmetrical relationships among global oil prices and local food prices.

183 **1. The panel linear ARDL Model**

184 Given the assumption that oil prices have a symmetric influence on food prices —the effect is
185 similar if oil prices increase or decrease —and referring to Salisu and Isah (2017), the following
186 formula is expressed by a symmetric form of linear ARDL:

$$\begin{aligned}
187 \quad \Delta(LFDP_t) = & \sum_{i=1}^q \beta_{1i} \cdot \Delta(LFDP_{t-i}) + \sum_{i=0}^p \beta_{2i} \cdot \Delta(LOLP_{t-i}) + \sum_{i=0}^p \beta_{3i} \cdot \Delta(LINF_{t-i}) + \\
188 \quad & \sum_{i=0}^p \beta_{4i} \cdot \Delta(LTO_{t-i}) + \sum_{i=0}^p \beta_{5i} \cdot \Delta(LURB_{t-i}) + \alpha_1 LFDP_{t-1} + \alpha_2 LOLP_{t-1} + \alpha_3 LINF_{t-1} + \\
189 \quad & \alpha_4 LTO_{t-1} + \alpha_5 LURB_{t-1} + \mu_i + \varepsilon_{it} \tag{1}
\end{aligned}$$

190 Where:

191 Δ is defined as the operator of differences; $i = 1, 2, \dots, N$ refers to each country's numbers; t
192 $= 1, 2, \dots, T$ denotes the time periods; p and q represent the optimum lag for dependent and
193 independent variables, respectively; and u_i is the group-specific effect. The residuals ε_{it} are
194 assumed to be white noise. $\beta_{1i}, \beta_{2i}, \beta_{3i}, \beta_{4i}, \beta_{5i}$ are the parameters for short-term while
195 $\alpha_1, \alpha_2, \alpha_3, \alpha_4, \alpha_5$ are the parameters for long-term.

196 To estimate short-term dynamic coefficients (e.g. Kun et al., 2015), the equation (1) has been
 197 re-specified by using the model of panel error correction model (ECM) as:

$$198 \quad \Delta(LFDP_t) = \sum_{i=1}^q \beta_{1i} \cdot \Delta(LFDP_{t-i}) + \sum_{i=0}^p \beta_{2i} \cdot \Delta(LOLP_{t-i}) + \sum_{i=0}^p \beta_{3i} \cdot \Delta(LINF_{t-i}) +$$

$$199 \quad \sum_{i=0}^p \beta_{4i} \cdot \Delta(LTO_{t-i}) + \sum_{i=0}^p \beta_{5i} \cdot \Delta(LURB_{t-i}) + \lambda_i ECT_{t-1} + \mu_i + \varepsilon_{it} \quad (2)$$

200 Where:

201 The co-integration term: $ECT_{t-1} = LFDP_{t-1} - \alpha_{1i}LOLP_{t-1} - \alpha_{2i}LINF_{t-1} - \alpha_{3i}LTO_{t-1} -$
 202 $\alpha_{4i}LURB_{t-1}$ is referred to the error correction term (ECT). λ_i is the coefficient of speed
 203 adjustment in error correction model towards a long-term equilibrium and is required to be
 204 significant and negative. $\beta_{1i}, \beta_{2i}, \beta_{3i}, \beta_{4i}, \beta_{5i}$ are short-run coefficients as shown in equation
 205 (1).

206 2. The panel non-linear ARDL Model

207 In contrast to the symmetric linear ARDL Model, asymmetries are calculated in this case to
 208 investigate food price asymmetric reactions to oil prices by decomposing the sum of negative
 209 and positive partial following Salisu and Isah (2017) and Shin et al. (2014). According to this
 210 model, positive and negative oil price shocks are not assumed to affect the price of food
 211 similarly. The ARDL model can therefore be expressed nonlinear in the following form:

$$212 \quad \Delta(LFDP_t) = \sum_{i=1}^q \beta_{1i} \cdot \Delta(LFDP_{t-i}) + \sum_{i=0}^p \beta_{2i}^+ \cdot \Delta(LOLP_{t-i}^+) + \sum_{i=0}^p \beta_{3i}^- \cdot \Delta(LOLP_{t-i}^-) +$$

$$213 \quad \sum_{i=0}^p \beta_{4i} \cdot \Delta(LINF_{t-i}) + \sum_{i=0}^p \beta_{5i} \cdot \Delta(LTO_{t-i}) + \sum_{i=0}^p \beta_{6i} \cdot \Delta(LURB_{t-i}) + \alpha_1 LFDP_{t-1} +$$

$$214 \quad \alpha_2 LOLP_{t-1}^+ + \alpha_3 LOLP_{t-1}^- + \alpha_4 LINF_{t-1} + \alpha_5 LTO_{t-1} + \alpha_6 LURB_{t-1} + \mu_i + \varepsilon_{it} \quad (3)$$

215 Where:

216 $LOLP^-$ and $LOLP^+$ represent the logarithm of partial sums for changes in negative and positive
 217 oil prices, indicating respectively negative and positive oil price shocks:

$$218 \quad LOLP_t^+ = \sum_{j=1}^t \Delta LOLP_{ij}^+ = \sum_{j=1}^t \max(\Delta LOLP_{ij}, 0) \quad (4)$$

$$219 \quad LOLP_t^- = \sum_{j=1}^t \Delta LOLP_{ij}^- = \sum_{j=1}^t \min(\Delta LOLP_{ij}, 0) \quad (5)$$

220
$$LOLP_t^+ = \sum_{j=1}^t \Delta LOLP_{ij}^+ = \sum_{j=1}^t \max(\Delta LOLP_{ij}, 0) \quad (4)$$

221
$$LOLP_t^- = \sum_{j=1}^t \Delta LOLP_{ij}^- = \sum_{j=1}^t \min(\Delta LOLP_{ij}, 0) \quad (5)$$

222 Since we added the error correction term of the linear model in Equation (1), we likewise utilize
 223 this term for the non-linear version as follows :

224
$$\Delta(LFDP_t) = \sum_{i=1}^q \beta_{1i} \cdot \Delta(LFDP_{t-i}) + \sum_{i=0}^p \beta_{2i}^+ \cdot \Delta(LOLP_{t-i}^+) + \sum_{i=0}^p \beta_{3i}^- \cdot$$

 225
$$\Delta(LOLP_{t-i}^-) + \sum_{i=0}^p \beta_{4i} \cdot \Delta(LINF_{t-i}) + \sum_{i=0}^p \beta_{5i} \cdot \Delta(LTO_{t-i}) + \sum_{i=0}^p \beta_{6i} \cdot \Delta(LURB_{t-i}) +$$

 226
$$\lambda_i ECT_{t-1} + \mu_i + \varepsilon_{it} \quad (6)$$

227 The error-correction term ECT_{t-1} captures the long-term equilibrium of the nonlinear panel
 228 ARDL is captured by the error-correction term ECT_{t-1} as equation (2). λ_i is an error-correction
 229 parameter that determines the speed at which the independent variable adjusts to reach its long-
 230 term equilibrium as a result of shocks in the dependent variable.

231 Results and discussions

232 1 Data Stationary Test Result

233 As a prerequisite for selecting an econometric model, panel unit root tests are applied on each
 234 variable to ensure that the data used are stationary at levels or first-order differences. With the
 235 chosen samples being countries, we suspend the cross-section dependence across them in our
 236 model. We utilized the cross-sectional dependence Pesaran CD test to assure this assumption.
 237 The CD-test findings indicate the presence of cross-sections in all variables, as shown in Table
 238 4, where the statistical values were significant at 1%, indicating that the sample countries share
 239 the same characteristics.

240 **Table 4** Cross-sectional dependence test

	Oil-exporting countries	Oil-importing countries	MENA countries
Variables	CD-test	CD-test	CD-test
<i>IFDP</i>	16.178*** (0.000)	16.842*** (0.000)	34.553*** (0.000)
<i>LOLP</i>	24.249*** (0.000)	17.748*** (0.000)	43.715*** (0.000)

<i>LINF</i>	6.418*** (0.000)	2.239** (0.025)	8.512*** (0.000)
<i>LTO</i>	3.984*** (0.000)	5.29*** (0.000)	8.683*** (0.000)
<i>LURB</i>	19.707*** (0.000)	9.505*** (0.000)	30.881*** (0.000)

241 Note. (1)The CD test is based upon the null hypothesis of the cross-section independence tends to $N(0,1)$.
242 A p-value near zero indicates the correlation between panel sets.

243 (2) Parentheses denote probability values, while ***,**,* represent a 1%, 5 and 10%, respectively, of
244 significance.If no special instructions, the following symbols are the same.

245 Given the cross-section dependencies among countries, first-generation panel unit root tests
246 such as Levin Lin and Chu (LLC), Im Pesaran and Shin (IPS) tests are unsuitable. Hence, we
247 employ the Cross-sectional Augmented Dickey-Fuller (CADF) Pesaran test (Pesaran, 2007) for
248 checking the stationarity levels of the variables in the sample countries. The findings of the
249 CADF tests indicate in Table 5 that the variables considered in our study are not integrated at
250 an order greater than I (1). These levels of integration confirm the convenience of the panel
251 ARDL approach.

252 **Table 5** CADF test results

Variables	Oil-exporting countries		Oil-importing countries		MENA countries	
	Zt-bar		Zt-bar		Zt-bar	
	Level	F.difference	Level	F.difference	level	F.difference
<i>LFDP</i>	-1.675** (0.047)	-2.722** (0.003)	0.861 (0.805)	-3.464*** (0.000)	-0.820 (0.206)	-3.049*** (0.001)
<i>LOLP²</i>	12.743 (1.000)	11.918 (1.000)	11.035 (1.000)	10.321 (1.000)	16.895 (1.000)	15.994 (1.000)
<i>LINF</i>	-0.945 (0.172)	-4.753*** (0.000)	0.344 (0.634)	-6.428 *** (0.000)	-1.159 (0.123)	-6.574 *** (0.000)
<i>LTO</i>	0.325 (0.627)	-3.029*** (0.001)	1.225 (0.890)	-1.891** (0.029)	1.547 (0.939)	-3.176*** (0.001)
<i>LURB</i>	1.101 (0.864)	-3.346** (0.002)	-2.317** (0.010)	3.049 (0.999)	-1.723** (0.042)	6.478 (1.000)

253 Note. The Pesaran's CADF-test is based upon the null hypothesis that series are integrated at I (1).

254 **2 Food prices symmetrical response to oil prices**

255 The estimated symmetric impact is summarized in Table 6 using equation (1). Here, all the
256 equations are estimated with pooled mean group (PMG) of dynamic heterogeneous panels
257 (Pesaran et al., 1999), Pesaran and Smith's (1995) mean group estimator (MG), and the

² Here the oil price variable is non-stationary because oil price only have the time change and no changes across countries.

258 Dynamic panel model with Fixed Effects (DFE). The Hausman test p-values for MG and DFE
 259 are not significant which indicates that the PMG is the adequate estimator in all sample
 260 countries for modelling the symmetric nexus among food prices and oil prices. Thus, we accept
 261 the estimated results for PMG methods.

262 **Table 6** Symmetric ARDL model results

Variables	Oil-exporting countries			Oil-importing countries			MENA countries		
	PMG	MG	DFE	PMG	MG	DFE	PMG	MG	DFE
Long-run									
<i>LOLP_{t-1}</i>	0.334*** (0.000)	-0.0758 (0.650)	0.341*** (0.000)	0.269*** (0.000)	0.166* (0.024)	-6.390 (0.721)	0.274*** (0.000)	0.027 (0.787)	1.652 (0.666)
<i>LINF_{t-1}</i>	0.0499** (0.012)	0.245 (0.092)	0.0414 (0.687)	0.0375* (0.062)	0.268* (0.072)	-6.011 (0.709)	0.0464** (0.002)	0.255** (0.011)	1.729 (0.679)
<i>LTO_{t-1}</i>	-0.357** (0.016)	0.295 (0.616)	0.175 (0.473)	-0.416*** (0.000)	-0.293 (0.177)	4.807 (0.805)	-0.369*** (0.000)	0.0430 (0.901)	-0.352 (0.908)
<i>LURB_{t-1}</i>	2.385*** (0.000)	145.9 (0.211)	1.163 (0.478)	2.035*** (0.000)	-4.971 (0.741)	40.37 (0.684)	2.081*** (0.000)	81.22 (0.234)	-10.36 (0.738)
<i>ECT</i>	-0.314* (0.062)	-0.529** (0.013)	-0.107*** (0.000)	-0.442*** (0.000)	0.167 (0.177)	0.00493 (0.666)	-0.244** (0.015)	-0.492*** (0.000)	-0.0111 (0.646)
Short-run									
$\Delta LOLP_{t-1}$	-0.0435* (0.014)	-0.00940 (0.715)	0.0105 (0.547)	-0.0370 (0.250)	-0.0659** (0.013)	-0.0382 (0.158)	-0.0386** (0.014)	-0.0336 * (0.085)	0.00553 (0.757)
$\Delta LINF_{t-1}$	-0.00183 (0.575)	-0.0295 (0.180)	-0.0005 (0.956)	0.00822 (0.276)	-0.0264* (0.070)	-0.0061 (0.318)	0.00324 (0.394)	-0.0282** (0.038)	-0.00350 (0.472)
ΔLTO_{t-1}	-0.317 (0.220)	-0.240 (0.089)	-0.101 (0.283)	0.0396 (0.639)	0.0419 (0.598)	0.0689 (0.403)	-0.125 (0.272)	-0.119 (0.200)	-0.0241 (0.714)
$\Delta LURB_{t-1}$	867.6 (0.319)	1168.1 (0.355)	-1.102** (0.027)	-16.53 (0.130)	1.673 (0.820)	-0.0417 (0.945)	387.7 (0.331)	668.2 (0.352)	0.0979 (0.844)
Hausman test		0.95 (0.9166)	5.08 (0.3695)		2.15 (0.7077)	0.68 (0.8593)		1.49 (0.8289)	2.35 (0.6582)
Model		PMG	PMG		PMG	PMG		PMG	PMG

263 Note. The PMG estimator is accepted under the null hypothesis of the Hausman test, while the MG estimator
 264 is accepted under the alternative hypothesis.

265 As shown in Table 6, the long-run finding shows that oil prices impact food prices positively
 266 and significantly for all samples. This corresponds to several empirical studies (e.g. Alghalith,
 267 2010; Baumeister and Kilian, 2014; Olayungbo and Hassan, 2016; Taghizadeh-hesary et al.,
 268 2019), suggesting that a rising oil price will induce higher food prices in the long-term. Further,
 269 we observe that the coefficients of *LOLP* of *oil-exporters* are greater than those for *oil-importers*
 270 which indicate the food prices of *oil-exporters* are rising more than that of *oil-importers* with
 271 an increasing oil price. The reason may be that the revenue gains from high oil prices for *oil-*
 272 *exporters* will spur more food demand and increase energy consumption costs in the food
 273 production process and ultimately lead to higher local food prices. It is also important to note
 274 that the total social welfare of *oil-exporters* may decrease because high food import costs offset

275 the benefits of oil exports. Another phenomenon is that oil prices have a significantly negative
276 short-term influence on *oil-exporters'* food prices, unlike for *oil-importers* where this is not
277 significant. The reason for this is that *oil-exporters* can take short-term measures to mitigate the
278 negative effects on the food system from high oil prices, while oil-importers have greater
279 difficulty in taking effective short-run measures.

280 The rate of inflation (INF) is similarly affected by food prices for both *oil-exporters* and *oil-*
281 *importers* positively and significantly in the long-term, while the short-term effect insignificant.
282 This finding is in line with Furceri et al. (2016), indicating that inflation has boosted food prices.
283 However, this effect is larger for *oil-exporters* than *oil-importers*. Specifically, a 1% increase
284 in the inflation rate increases food prices in the long-term by 0.049% for *oil-exporters* while
285 this change is 0.037% for the oil-importing group in the long-term. This indicates that prices of
286 food for *oil-exporters* are more sensitive than for *importers*.

287 In addition, the trade openness index (TO) affected negatively and significantly the long-run
288 food prices for all samples .This effect can be explained by the competitiveness of food
289 commodities when MENA countries open up to foreign trade, leading to the further long-term
290 declines in food prices. The short-term coefficient of trade openness was found to be negative
291 and insignificant in the MENA region. The reason may be that it takes a long time for the
292 welfare effect of trade freedom to impact food prices.

293 The long-run results show that urban population (URB) also affects food prices positively and
294 is statistically significant for the sampled countries. This implies that rural migration to cities
295 within the same country—or between MENA countries— this has resulted in a substantial
296 increase in the urban population thereby depriving labour from the agricultural sector in rural
297 areas. As a consequence, the agricultural sector's performance in these areas has deteriorated,
298 leading to higher food prices.

299 For all sample countries, the error correction coefficients (ECTs) terms are statistically
300 significant, negative, and less than one, demonstrating that short-run fluctuations in the system
301 will converge in a long-run relationship. For oil-exporters, the distortions due to food price
302 shocks can be corrected at a speed of 31.4%. Alternatively, for oil-importers, the speed of long-
303 run equilibrium adjustment is 44.2%— which is faster than for the oil-exporting group.

304 To test how robust the estimated outcomes are, this study re-estimated the PMG model by
305 gradually adding variables and the panel OLS results are reported (see Appendix Table A, B,
306 and C). The former results show that as the number of variables increases, the coefficients of
307 most variables fluctuated in a small range, verifying that the estimation results are robust. The
308 latter results indicate that, as expected, most of the variables are correctly signed, proving that
309 the long-run result of PMG estimates is reasonable.

310 **3 Food prices asymmetrical response to oil prices**

311 Food prices in MENA countries may have different impact mechanisms in the face of
312 fluctuating oil prices. To explore this asymmetric response, we use panel data samples for
313 country groups and time series for each country to analyse the heterogeneous response of
314 different samples.

315 **(1) Estimation results using panel data**

316 With the null hypothesis that positive and negative changes are not jointly significant, Wald
317 tests are conducted to investigate the existence of an asymmetry relationship with panel data
318 used for both groups of *oil-exporters* and *oil-importers* nations. Table 7 shows that for the entire
319 sample of MENA countries, the null hypothesis is rejected for both short- and long-term, which
320 supports the asymmetrical relationship and implies that oil prices do not have the identical
321 impact on food price as oil prices rise and decrease. From the perspective of country groups,
322 however, the F-statistics of the Wald test are not significant. This result predicts the
323 heterogeneous response within the oil importing and exporting countries groups.

324 **Table 7** Wald Test for Asymmetry

	Oil-exporting countries		Oil-importing countries		MENA countries	
	Long-run	Short-run	Long-run	Short-run	Long-run	Short-run
Wald Statistic	0.95 (0.3301)	1.52 (0.2180)	0.43 (0.5126)	1.51 (0.2190)	3.25* (0.0713)	6.38** (0.0115)
Asymmetric Evidence	No	No	No	No	Yes	Yes

325 Note. The null hypothesis of Wald's test that positive and negative changes are not jointly significant. Thus,
326 they do not differ from zero, the alternative is that the changes are jointly significant.

327 The estimated asymmetric impacts of changes in oil price on food price sample groups is
328 presented in Table 8 using equation (3). First, all equations are estimated using PMG, MG, and
329 DFE estimators, and then choose the most appropriate estimator based on the Hausman test in
330 the last line of Table 8. The findings show that the null hypothesis is appropriate, which suggests
331 that all groups are consistent with the PMG estimator.

332 **Table 8** Estimation of NARDL Models.

Variables	Oil-exporting countries	Oil-importing countries	MENA countries
Long-run			
$LOLP^-_{t-1}$	0.373*** (0.000)	0.287*** (0.000)	0.296*** (0.000)
$LOLP^+_{t-1}$	0.354*** (0.000)	0.301*** (0.000)	0.302*** (0.000)
$LINF_{t-1}$	0.0688** (0.003)	0.0373** (0.005)	0.0256 (0.151)
LTO_{t-1}	-0.361** (0.032)	-0.358*** (0.000)	-0.408*** (0.000)
$LURB_{t-1}$	2.354*** (0.000)	1.992*** (0.000)	1.915*** (0.000)
ECT	-0.262* (0.052)	-0.175 (0.171)	-0.250* (0.025)
Short-run			
$\Delta LOLP^+_{t-1}$	-0.0229* (0.051)	-0.0382 (0.326)	-0.0494** (0.021)
$\Delta LOLP^-_{t-1}$	-0.0255* (0.057)	-0.0360 (0.365)	-0.0470** (0.027)
$\Delta LINF_{t-1}$	-0.00333 (0.458)	0.0140 (0.220)	0.00624 (0.241)
ΔLTO_{t-1}	-0.257 (0.141)	0.0439 (0.619)	-0.183 (0.265)
$\Delta LURB_{t-1}$	754.2 (0.317)	-18.67* (0.097)	460.0 (0.328)
Hausman test	0.37 (0.9961)	3.63 (0.6046)	0.55 (0.2303)
Model	PMG	PMG	PMG

333 For the MENA sample, the short-term asymmetrical parameters of $\Delta LOLP^+_{t-1}$ and
334 $\Delta LOLP^-_{t-1}$ are negative at 5% level of significance, implying that food prices will decline over

335 the short-term, whether oil prices increase or decrease. One possible reason is that in the short-
336 term, governments provide timely food subsidies and other measures to stabilize prices.
337 However, the long-term asymmetric parameters (0.296 and 0.302) are positively significant at
338 the 1% level, which indicates, whether oil prices rise or fall, food prices always rise. This
339 confirms that continuing food price rises is inevitable over the long-term even though it will
340 decline in the short-term. Another observed phenomenon is that when oil prices rise, food prices
341 rise faster than the corresponding decline when oil prices fall. This finding indicates that when
342 the international oil price rises, it exerts more pressure on the domestic food price.
343 Error correction coefficients (ECTs) terms are negatively significant for the samples of oil-
344 exporting and MENA countries that support a long-term convergence. When a short-term
345 deviation is caused by shocks in food prices, the adjustment to the long-run equilibrium in the
346 long-term is 26.2% and 25% in oil-exporting and MENA countries, respectively. Alternatively,
347 the coefficient of ECT for oil-importing countries is shown to be negative but statistically
348 insignificant indicating no convergence in the long-run relationship.

349 (2) Estimation results using time series

350 We estimate the non-linear effects of oil price on each country's food price by employing time
351 series for the two groups (oil-exporters and oil-importers) following the non-linear ARDL
352 model. The Wald test results for asymmetries are summarized in Table 9.

353 **Table 9** Wald Test for Asymmetry in MENA countries

Country	Wald Statistic		Asymmetric Evidence	
	Long-run	Short-run	Long-run	Short-run
Oil-exporting countries				
Algeria	48.39*** (0.000)	0.07652 (0.788)	Yes	No
Bahrain	32.93*** (0.000)	0.3956 (0.543)	Yes	No
Iraq	15.78** (0.003)	2.032 (0.184)	Yes	No
Kuwait	2.148 (0.173)	7.584 (0.979)	No	No
Oman	13.72** (0.004)	0.0148 (0.906)	Yes	No
Qatar	7.02** (0.024)	0.8604 (0.375)	Yes	No

Saudi Arabia	0.08213 (0.780)	0.4131 (0.535)	No	No
UAE	0.7371 (0.411)	1.64 (0.229)	No	No
<hr/>				
Oil-importing countries				
Egypt	529.5*** (0.000)	6.211** (0.032)	Yes	Yes
Jordan	31.55*** (0.000)	0.2005 (0.664)	Yes	No
Lebanon	57.7*** (0.000)	1.31 (0.279)	Yes	No
Morocco	87.09*** (0.000)	0.0010 (0.975)	Yes	No
Tunisia	0.2471 (0.630)	0.2482 (0.629)	No	No
West Bank & Gaza	8.194** (0.017)	0.3595 (0.562)	Yes	No

354 The asymmetry test findings for oil-exporting group—including Algeria, Bahrain, Iraq, Oman,
355 and Qatar—indicated that the Wald test F-statistics are significant in the long-term, suggesting
356 that asymmetrical influence exists in those countries. Alternatively, economies such as Kuwait,
357 Saudi Arabia, and the United Arab Emirates have rejected the null hypothesis regarding the
358 existence of asymmetric linkage in the short- and long-term, which indicates that the oil-food
359 prices relationship is not asymmetric. This evidence supports our observations of the cumulative
360 effects of oil price on food price (see Appendix Figure A), where the asymmetry line shows
361 that food prices in Kuwait, Saudi Arabia, and the UAE do not react differently to shocks in the
362 oil price—whether increasing or decreasing.

363 Except for Tunisia, all oil-importers pass the long-term Wald test at 5% significance level,
364 indicating how asymmetric the changes in the price of oil are influencing long-term food price.
365 The asymmetry line implies that the impact of oil price on food price in Egypt, Jordan, Lebanon,
366 Morocco, the West Bank, and Gaza is not identical for either a rise or decline in oil price, as
367 seen in the plots of the cumulative effects of oil price on food price (see Appendix Figure B).

368 The long-run asymmetric effects for each MENA country are shown in Table 10. For the oil-
369 exporting group, the results show that the coefficients associated with increases in the oil price
370 positively affect food prices. Specifically, an increase of 1% in positive oil price changes causes

371 an increase of between 0.24% and 0.59% in food prices. Nevertheless, the coefficients related
 372 to decreases in the oil price negatively affect food prices in Iraq, whereas in Algeria, Bahrain,
 373 Oman, and Qatar these are insignificant. For oil-importing nations, the outcomes also indicate
 374 that food prices are positively influenced by the increasing oil price changes. A 1% rise in oil
 375 prices results in a 0.12% to 0.40% increase in food prices. However, the coefficients of oil price
 376 reductions have been found to be mixed. Egypt and Lebanon are positively affected, while
 377 Jordan is negatively affected, and the impact on Morocco, the West Bank and Gaza are
 378 insignificant.

379 **Table 10** Long-run asymmetric effects

Oil exporting countries								
	Algeria	Bahrain	Iraq	Kuwait	Oman	Qatar	Saudi Arabia	UAE
$LOLP^{+}_{t-1}$	0.259** (0.014)	0.239** (0.039)	0.586*** (0.000)	-	0.379*** (0.000)	0.338** (0.003)	-	-
$LOLP^{-}_{t-1}$	0.148 (0.291)	-0.015 (0.914)	-0.309** (0.009)	-	-0.161 (0.201)	-0.159 (0.283)	-	-
Oil importing countries								
	Egypt	Jordan	Lebanon	Morocco	Tunisia	West Bank & Gaza		
$LOLP^{+}_{t-1}$	0.255*** (0.001)	0.403*** (0.000)	0.214** (0.003)	0.124*** (0.000)	-	0.298*** (0.000)		
$LOLP^{-}_{t-1}$	0.399*** (0.000)	-0.173* (0.067)	0.153* (0.097)	0.039 (0.296)	-	-0.143 (0.140)		

380 By comparing our time-series results with estimates of the panel data, we concluded that due to
 381 countries' different economic status, the asymmetric impacts of fluctuations in oil prices on the
 382 food price varies. In general, our conclusion is that when oil prices rise, the food prices also
 383 rise; but when oil prices drop, food prices do not always decrease. This also proves that, as a
 384 necessary commodity, the price of food is sticky—it does not fall easily.

385 **Conclusions and policy implications**

386 The linear and nonlinear panel ARDL models are used in this paper to investigate the
 387 symmetrical and asymmetrical relationship between world oil prices and local food prices for
 388 the MENA region from 2000 to 2020. Findings from the symmetric linear ARDL model show
 389 that oil prices have a long-run positive and significant impact on food prices for oil-exporting

390 and importing MENA nations. The positive impact is larger for *oil-exporters* than for *oil-*
391 *importers*. This outcome concludes oil prices have a greater influence on increases in food
392 prices for those economies that export oil at high prices. This results in increased costs for
393 imported food due to the higher energy cost needed to produce food commodities in the country
394 of origin. Furthermore, we found that the inflation rate and urban population affect positively
395 and significantly on food prices for oil-exporters and importers groups, while trade openness is
396 negative and significant in relation to food prices. These results suggest that expanding trade
397 openness can help these countries source cheaper food resources. Additionally, reducing
398 inflation and controlling the scale of urbanization will foster agricultural development and
399 ensure an adequate food supply.

400 For the entire sample from the MENA region, outcomes indicate that the oil price effect on food
401 price is asymmetrical in the short- and long-term although this effect is insignificant for the oil-
402 exporters and importers groups. This finding follows Meyer et al. (2018) and Ibrahim (2015).
403 However, the short-term asymmetric effect was negative, while this was found to be positive in
404 the long-term, indicating that—whether oil prices rise or fall—food prices always rise. These
405 results indicate that food products price is sticky: once food prices rise, it is harder to reduce
406 them. Therefore, policymakers should take measures to improve agricultural labour
407 productivity, increase the supply capacity of agricultural products, and develop renewable
408 energy sources (such as photovoltaic cells) to eliminate the dependence of agriculture on oil.
409 For the MENA region, developing biomass energy is discouraged to prevent the reduction in
410 available agricultural land. Thus, while developing biomass energy, the government—should
411 reasonably assess its potential impact on agricultural land and food supply.

412 Regarding the non-linear ARDL results for each country using time series, we found an absence
413 of asymmetrical behaviour for nations including Kuwait, Saudi Arabia, and the UAE from the
414 oil-exporting group, and Tunisia from the oil-importing group. Except those countries, others

415 have the asymmetrical effects in food prices responses to oil prices in the long-term. Oil prices
416 have a major impact on countries such as Iraq and Jordan, and food price tracked oil price
417 fluctuations. To maintain the stability of food prices, these countries need to make counter-
418 cyclic adjustments; when the oil price is expected to rise, they should stockpile a large amount
419 of food, and when the oil price is expected to fall, they should sell food. For countries such as
420 Egypt and Lebanon, The rate of increase in food prices fluctuates with rising oil prices; hence,
421 these governments need to establish long-term food price stability mechanisms—increasing
422 food production, establishing a stable international food trading partner and other policies
423 regardless of how oil prices fluctuate. Additionally, for countries such as Algeria and Bahrain
424 which are seriously affected by increases in the oil price—but not by its decline—they need to
425 stockpile more agricultural products before forecasting higher oil prices.

426 As one of the global regions with insufficient food supply, food price stability in the MENA
427 region is an important guarantee of food security. First, MENA governments should review
428 their agricultural policies by providing incentives and implementing effective mechanisms to
429 increase domestic production to avoid the effects of high food prices that result from energy
430 price fluctuations; second, as most MENA agricultural resources (whether water or arable land)
431 are located in oil-importing countries (characterized by a scarcity of financial resources) while
432 *oil-exporters* have enormous oil wealth (matched by a scarcity of arable land and water)
433 cooperation should be strengthened between countries to stabilize oil and food prices. Third,
434 according to the differentiated symmetrical and asymmetrical mechanisms of food and oil
435 prices, MENA countries should establish specific coping strategies. For countries with a
436 relatively large impact on international oil prices, an early warning mechanism could mitigate
437 price spikes in food and oil.

438 **Appendix:**

439 **Table A** The estimated results of the pooled mean group and panel OLS for oil-exporting countries

Pooled mean group

Panel OLS

Long-run					
<i>LOLP</i> _{<i>t-1</i>}	0.578*** (0.000)	0.498*** (0.000)	0.452*** (0.000)	0.334*** (0.000)	0.283** (0.008)
<i>LINF</i> _{<i>t-1</i>}		0.0929* (0.088)	0.112** (0.037)	0.0499** (0.012)	-0.0541** (0.023)
<i>LTO</i> _{<i>t-1</i>}			0.0431 (0.876)	-0.357** (0.016)	-0.311 (0.335)
<i>LURB</i> _{<i>t-1</i>}				2.385*** (0.000)	3.502** (0.002)
ECT	-0.109*** (0.000)	-0.0864*** (0.000)	-0.0833*** (0.000)	-0.314* (0.062)	
Short -run					
Δ . <i>LOLP</i> _{<i>t-1</i>}	-0.0135 (0.632)	-0.00907 (0.721)	-0.00223 (0.870)	-0.0435** (0.014)	
Δ . <i>LINF</i> _{<i>t-1</i>}		0.00299 (0.622)	0.00122 (0.821)	-0.00183 (0.575)	
Δ . <i>LTO</i> _{<i>t-1</i>}			-0.0616 (0.319)	-0.317 (0.220)	
Δ . <i>LURB</i> _{<i>t-1</i>}				867.6 (0.319)	

440

441 **Table B** The estimated results of the pooled mean group and panel OLS for oil-importing countries

	Pooled mean group				Panel OLS ⁴⁴²	
Long-run						
<i>LOLP</i> _{<i>t-1</i>}	0.594*** (0.000)	0.738** (0.039)	0.486*** (0.000)	0.269*** (0.000)	0.321** (0.022)	443 444
<i>LINF</i> _{<i>t-1</i>}		1.633** (0.002)	0.0774* (0.064)	0.0375* (0.062)	0.139 (0.332)	445 446
<i>LTO</i> _{<i>t-1</i>}			-0.825*** (0.000)	-0.416*** (0.000)	-0.841 (0.106)	447
<i>LURB</i> _{<i>t-1</i>}				2.035*** (0.000)	2.874 (0.112)	448
ECT	-0.0566* (0.048)	- 0.0241*** (0.001)	-0.0606 (0.423)	-0.167 (0.177)		
Short-run						
Δ . <i>LOLP</i> _{<i>t-1</i>}	-0.0277* (0.069)	-0.0416* (0.050)	-0.0465 (0.102)	-0.0370 (0.250)		
Δ . <i>LINF</i> _{<i>t-1</i>}		-0.00128 (0.881)	0.0146 (0.241)	0.00822 (0.276)		
Δ . <i>LTO</i> _{<i>t-1</i>}			0.0200 (0.790)	0.0396 (0.639)		
Δ . <i>LURB</i> _{<i>t-1</i>}				-16.53 (0.130)		

449 **Table C** The estimated results of the pooled mean group and panel OLS for MENA countries

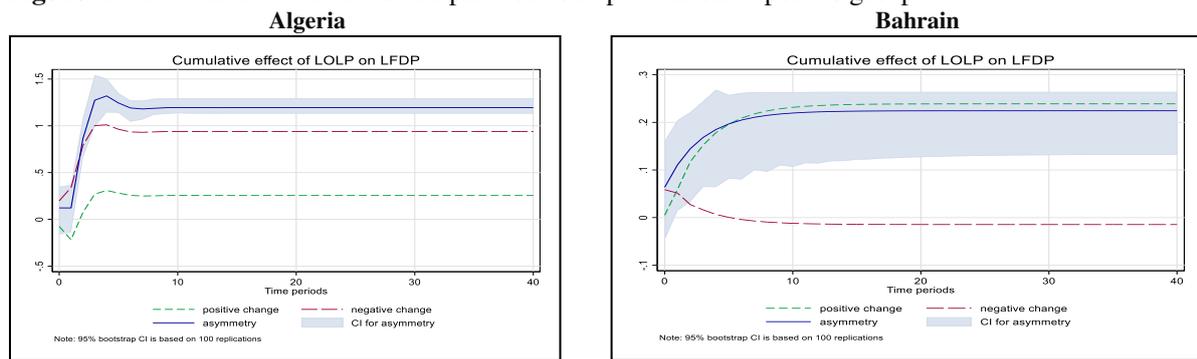
	Pooled mean group				panel OLS
Long-run					
<i>LOLP</i> _{<i>t-1</i>}	0.584*** (0.000)	0.525*** (0.000)	0.524*** (0.000)	0.274*** (0.000)	0.295*** (0.000)
<i>LINF</i> _{<i>t-1</i>}		0.122** (0.016)	0.176** (0.005)	0.0464** (0.002)	0.00979 (0.835)
<i>LTO</i> _{<i>t-1</i>}			-0.0624 (0.792)	-0.369*** (0.000)	-0.626* (0.051)
<i>LURB</i> _{<i>t-1</i>}				2.081*** (0.000)	3.227*** (0.000)

<i>ECT</i>	-0.0863*** (0.000)	-0.0636*** (0.000)	-0.0572*** (0.000)	-0.244** (0.015)
Short -run				
$\Delta.LOLP_{t-1}$	-0.0196 (0.247)	-0.0212 (0.195)	-0.0156 (0.250)	-0.0386** (0.014)
$\Delta.LINF_{t-1}$		0.00730 (0.201)	0.00579 (0.328)	0.00324 (0.394)
$\Delta.LTO_{t-1}$			-0.0393 (0.307)	-0.125 (0.272)
$\Delta.LURB_{t-1}$				387.7 (0.331)

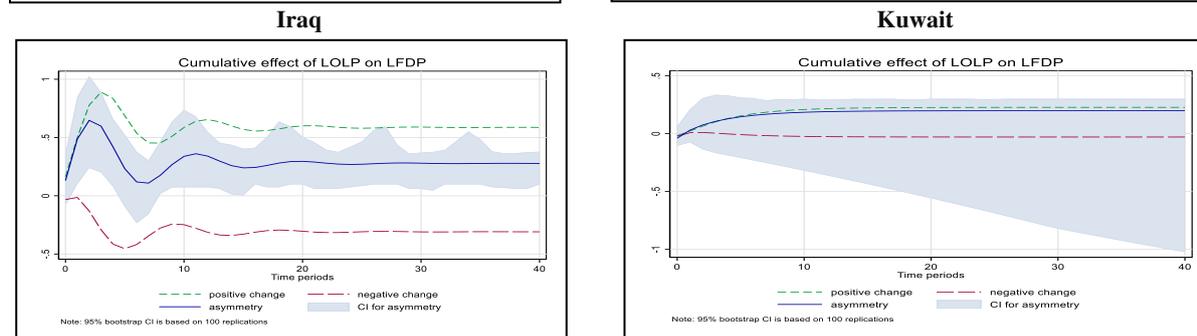
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Figure A The cumulative effect of oil price on food price in oil-exporters group

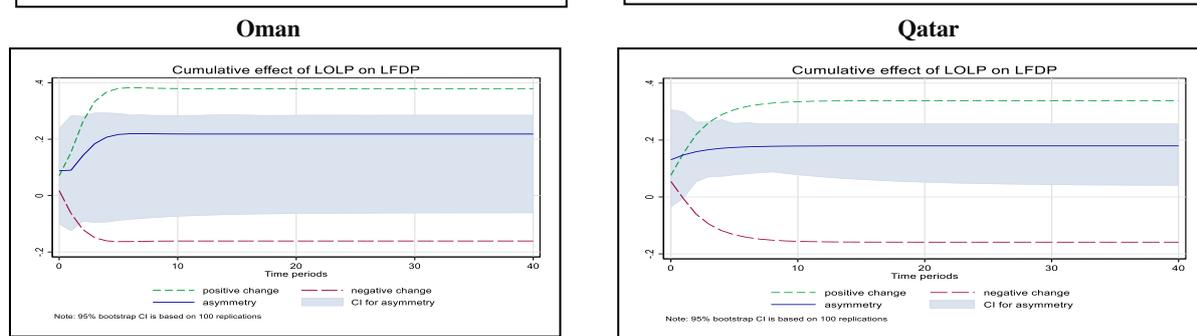
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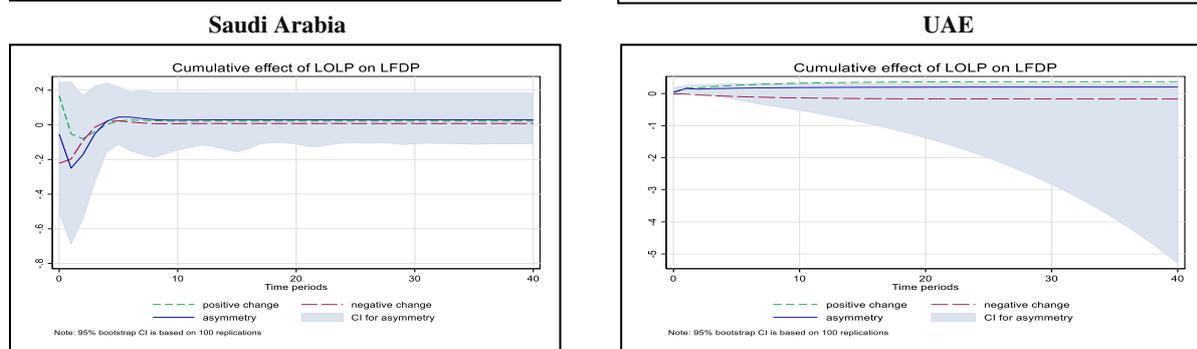
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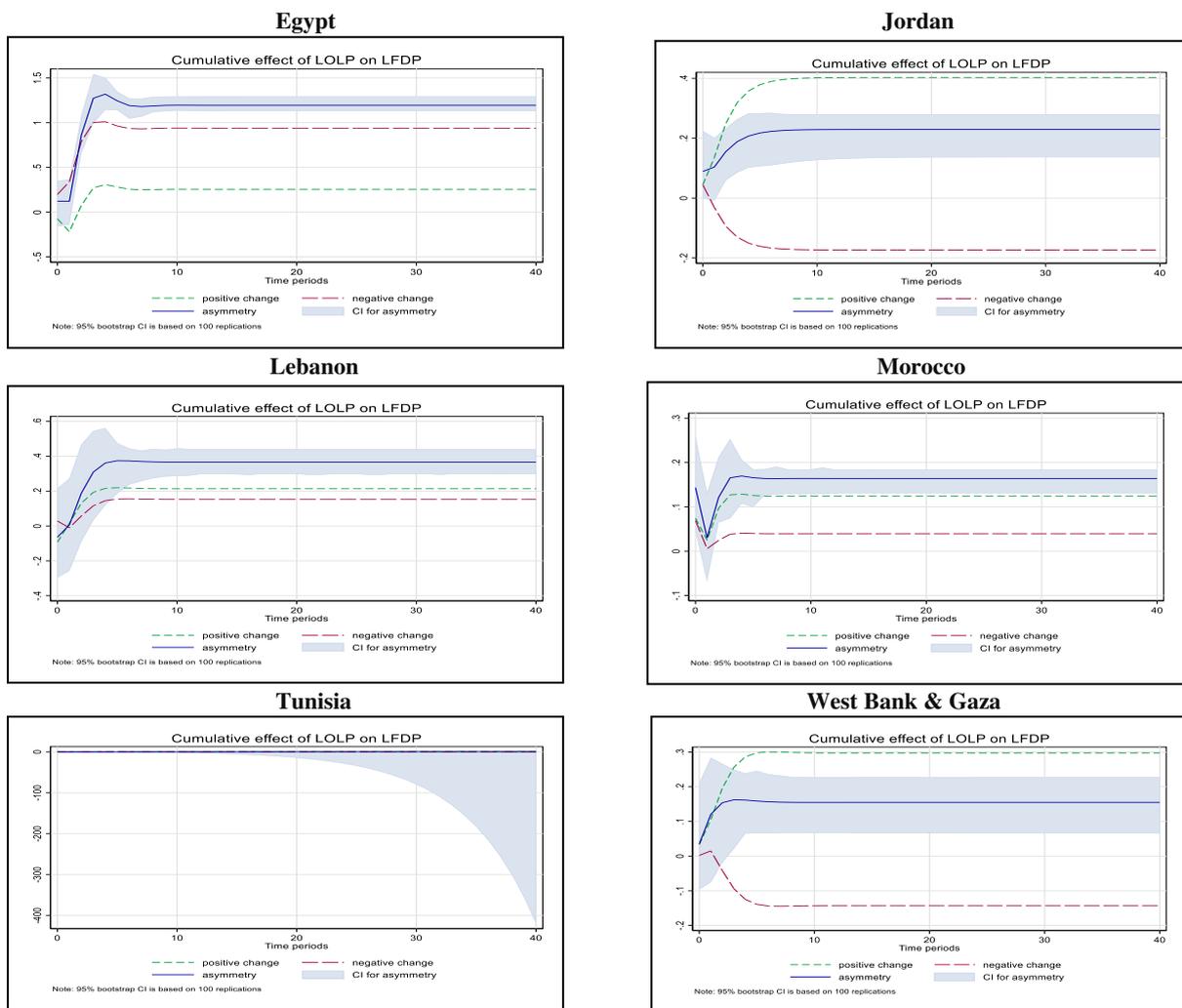


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459 **Figure B** The cumulative effect of oil price on food price in oil-importers group

460



465

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477 Writing, review and editing: Hadj Cherif Houda, Zhenling Chen;
478 Final approval of the article: Zhenling Chen; Guohua Ni
479 Overall responsibility: Hadj Cherif Houda, Zhenling Chen, Guohua Ni.
480 Zhenling Chen, Guohua Ni contributed equally to this work and share the corresponding author.

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482 All data generated or analyzed during this study are included in this published article.

483 **Compliance with ethical standards**

484 **Ethical approval:** Not applicable.

485 **Consent to participate:** Not applicable.

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488 **Competing interests:** The authors declare that they have no competing interest.

489

490 **References:**

491 Alghalith, M., 2010. The interaction between food prices and oil prices. *Energy Econ.* 32,
492 1520–1522. <https://doi.org/10.1016/j.eneco.2010.08.012>

493 Apergis, N., El-Montasser, G., Sekyere, E., Ajmi, A.N., Gupta, R., 2014. Dutch disease effect
494 of oil rents on agriculture value added in Middle East and North African (MENA)
495 countries. *Energy Econ.* 45, 485–490. <https://doi.org/10.1016/j.eneco.2014.07.025>

496 Baumeister, C., Kilian, L., 2014. Do oil price increases cause higher food prices ? *Econ.*
497 *Policy* 29, 691–774. <https://doi.org/https://doi.org/10.1111/1468-0327.12039>

498 Bellakhal, R., Ben Kheder, S., Haffoudhi, H., 2019. Governance and renewable energy
499 investment in MENA countries:How does trade matter? *Energy Econ.* 84, 104541.
500 <https://doi.org/10.1016/j.eneco.2019.104541>

501 Chen, D., Gummi, U.M., Umar, A.S., 2019. The review on the link between food and oil
502 markets in the view of price dynamics. *Am. J. Ind. Bus. Manag.* 09, 1890–1900.
503 <https://doi.org/10.4236/ajibm.2019.99122>

504 Chen, S., Kuo, H., Chen, C., 2010. Modeling the relationship between the oil price and global
505 food prices. *Appl. Energy* 87, 2517–2525.
506 <https://doi.org/10.1016/j.apenergy.2010.02.020>

507 Ciaian, P., Kancs, A., 2011. Interdependencies in the energy-bioenergy-food price systems:
508 A cointegration analysis. *Resour. Energy Econ.* 33, 326–348.
509 <https://doi.org/10.1016/j.reseneeco.2010.07.004>

510 Dalheimer, B., Herwartz, H., Lange, A., 2021. The threat of oil market turmoils to food price
511 stability in Sub-Saharan Africa. *Energy Econ.* 93, 105029.
512 <https://doi.org/https://doi.org/10.1016/j.eneco.2020.105029>

513 Ek Fälth, H., Atsmon, D., Reichenberg, L., Verendel, V., 2021. MENA compared to Europe:
514 The influence of land use, nuclear power, and transmission expansion on renewable
515 electricity system costs. *Energy Strateg. Rev.* 33, 100590.
516 <https://doi.org/10.1016/j.esr.2020.100590>

517 Engle, R.F., Granger, C.W.J., 1987. Co-integration and error
518 correction:Representationestimation,and testing. *Econometrica* 55, 251–276.

519 Esmaeili, A., Shokoohi, Z., 2011. Assessing the effect of oil price on world food prices:
520 Application of principal component analysis. *Energy Policy* 39, 1022–1025.
521 <https://doi.org/10.1016/j.enpol.2010.11.004>

522 Furceri, D., Loungani, P., Simon, J., Wachter, S.M., 2016. Global food prices and domestic
523 inflation: Some cross-country evidence. *Oxf. Econ. Pap.* 68, 665–687.
524 <https://doi.org/10.1093/oep/gpw016>

525 Gardebroek, C., Hernandez, M.A., 2013. Do energy prices stimulate food price volatility ?
526 Examining volatility transmission between US oil , ethanol and corn markets. *Energy*
527 *Econ.* 40, 119–129. <https://doi.org/10.1016/j.eneco.2013.06.013>

528 Gohin, A., Chantret, F., 2010. The long-run impact of energy prices on world agricultural
529 markets : The role of macro-economic linkages. *Energy Policy* 38, 333–339.
530 <https://doi.org/10.1016/j.enpol.2009.09.023>

531 Ibrahim, M.H., 2015. Oil and food prices in Malaysia: A nonlinear ARDL analysis. *Agric.*
532 *Food Econ.* 3, 0–14. <https://doi.org/10.1186/s40100-014-0020-3>

533 Johansen, S., Juselius, K., 1992. Testing structural hypotheses in a multivariate cointegration
534 analysis of the PPP and the UIP for UK. *J. Econom.* 53, 211–244.

535 Kassouri, Y., Altıntaş, H., 2020. Human well-being versus ecological footprint in MENA
536 countries: A trade-off? *J. Environ. Manage.* 263, 110405.
537 <https://doi.org/10.1016/j.jenvman.2020.110405>

538 Kun, S., Qi, X., Nee, Y., 2015. A comparative study on the effects of oil price changes on
539 inflation. *Procedia Econ. Financ.* 26, 630–636. [https://doi.org/10.1016/S2212-](https://doi.org/10.1016/S2212-5671(15)00800-X)
540 [5671\(15\)00800-X](https://doi.org/10.1016/S2212-5671(15)00800-X)

541 Meyer, D.F., Sanusi, K.A., Hassan, A., 2018. Analysis of the asymmetric impacts of oil prices
542 on food prices in oil-exporting developing countries. *J. Int. Stud.* 11, 82–94.
543 <https://doi.org/10.14254/2071-8330.2018/11-3/7>

544 Mokni, K., Ben-Salha, O., 2020. Asymmetric causality in quantiles analysis of the oil-food
545 nexus since the 1960s. *Resour. Policy* 69, 101874.
546 <https://doi.org/https://doi.org/10.1016/j.resourpol.2020.101874>

547 Nwoko, I.C., Aye, G.C., Asogwa, B.C., 2016. Oil price and food price volatility dynamics:
548 The case of Nigeria. *Cogent Food Agric.* 2, 1142413.
549 <https://doi.org/10.1080/23311932.2016.1142413>

550 Olayungbo, D., Hassan, W., 2016. Effects of oil price on food prices in developing oil
551 exporting countries: a panel autoregressive distributed lag analysis. *OPEC Energy Rev.*
552 40, 397–411. <https://doi.org/10.1111/opec.12090>

553 Pal, D., Mitra, S.K., 2018. Interdependence between crude oil and world food prices: A
554 detrended cross correlation analysis. *Phys. A Stat. Mech. its Appl.* 492, 1032–1044.
555 <https://doi.org/10.1016/j.physa.2017.11.033>

556 Pal, D., Mitra, S.K., 2016. Asymmetric oil product pricing in India: Evidence from a multiple
557 threshold nonlinear ARDL model. *Econ. Model.* 59, 314–328.
558 <https://doi.org/https://doi.org/10.1016/j.econmod.2016.08.003>

559 Pesaran, M.H., 2007. A simple panel unit root test in the Presence of cross-section
560 dependence. *J. Appl. Econom.* 312, 265–312. <https://doi.org/10.1002/jae>

561 Pesaran, M.H., Shin, Y., Smith, R.J., 2001. Bounds testing approaches to the analysis of level
562 relationships. *J. Appl. Econom.* 16, 289–326.

563 Pesaran, M.H., Shin, Y., Smith, R.P., 1999. Pooled mean group estimation of dynamic
564 heterogeneous panels. *J. Am. Stat. Assoc.* 94, 621–643.

565 Pesaran, M.H., Smith, R., 1995. Estimating long-run relationships from dynamic
566 heterogeneous panels. *J. Econom.* 68, 79–113.

567 Reboredo, J.C., 2012. Do food and oil prices co-move? *Energy Policy* 49, 456–467.
568 <https://doi.org/10.1016/j.enpol.2012.06.035>

569 Salisu, A.A., Isah, K.O., 2017. Revisiting the oil price and stock market nexus : A nonlinear
570 Panel ARDL approach. *Econ. Model.* 66, 258–271.
571 <https://doi.org/10.1016/j.econmod.2017.07.010>

572 Sarwar, M.N., Hussain, H., Maqbool, M.B., 2020. Pass through effects of oil price on food
573 and non-food prices in Pakistan: A nonlinear ARDL approach. *Resour. Policy* 69,
574 101876. <https://doi.org/10.1016/j.resourpol.2020.101876>

575 Shin, Y., Yu, B., Greenwood-Nimmo, M., 2014. Modelling asymmetric cointegration and
576 dynamic multipliers in a nonlinear ARDL framework. *Festschrift Honor Peter Schmidt*
577 44, 281–314. <https://doi.org/10.2139/ssrn.1807745>

578 Taghizadeh-hesary, F., Rasoulinezhad, E., Yoshino, N., 2019. Energy and food security :
579 Linkages through price volatility. *Energy Policy* 128, 796–806.

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Figures

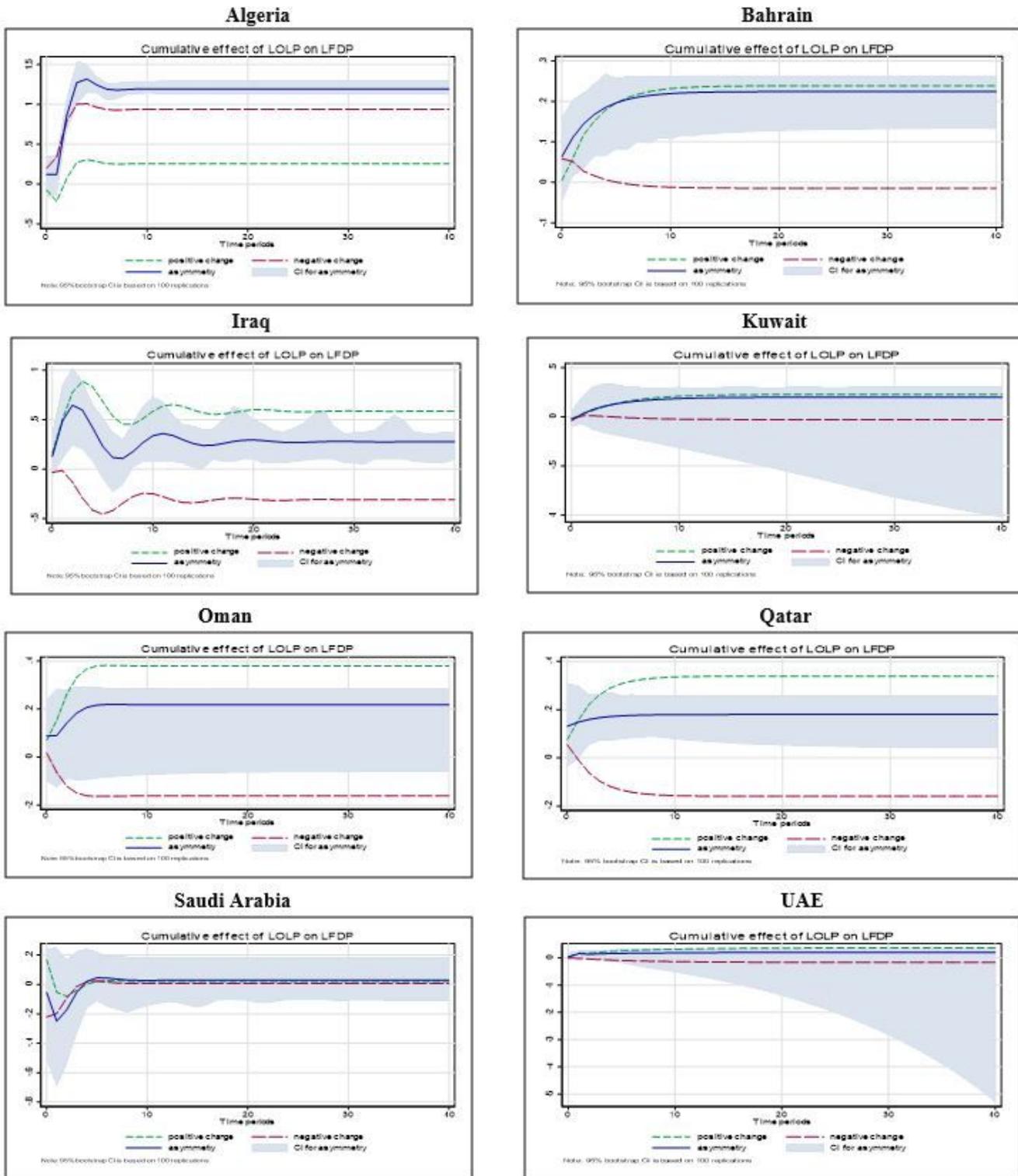


Figure 1

The cumulative effect of oil price on food price in oil-exporters group

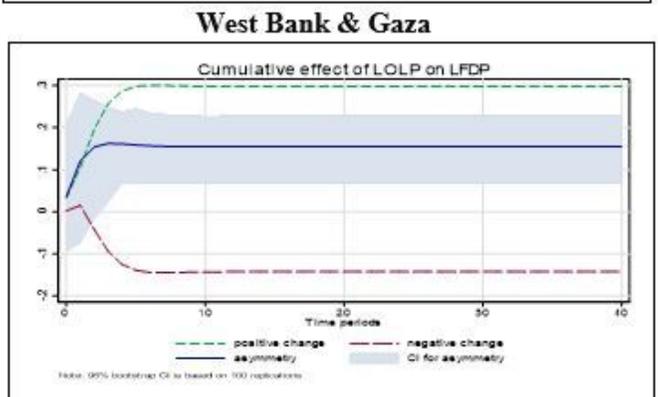
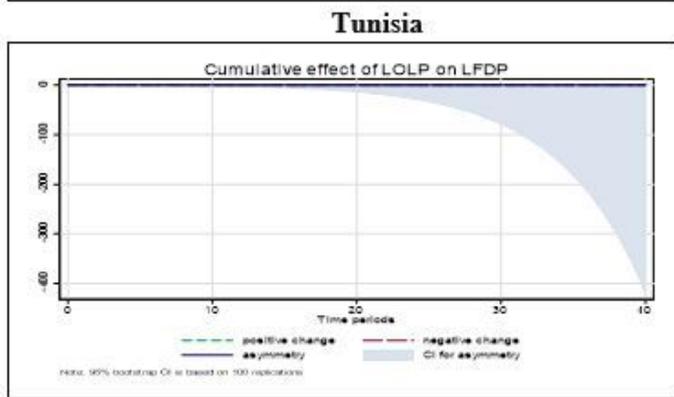
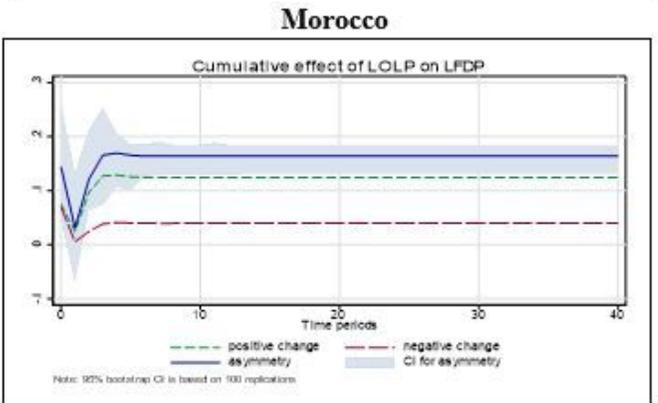
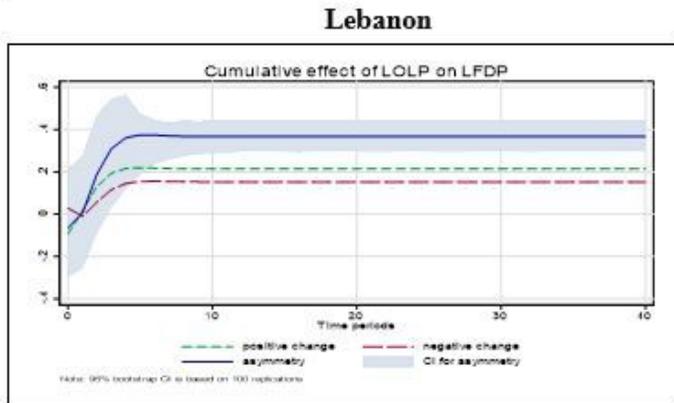
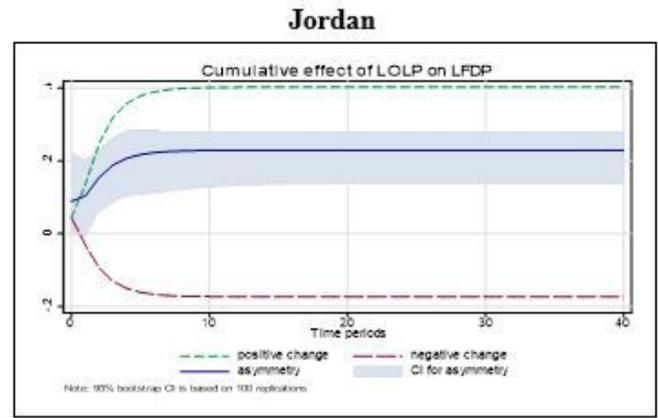
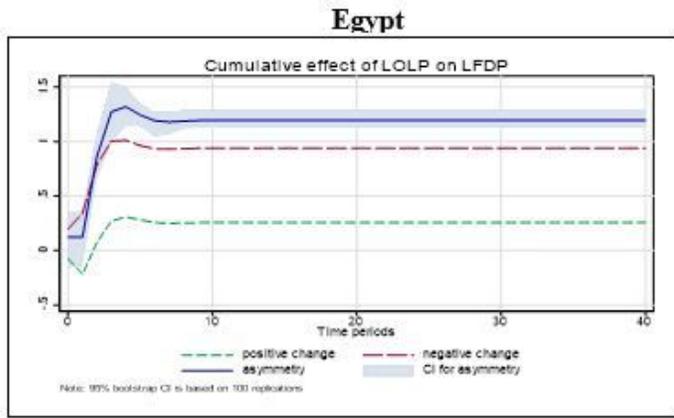


Figure 2

The cumulative effect of oil price on food price in oil-importers group