

Blue economy and economic development: The function of aquaculture and capture fisheries in Next-11 nations

Mavi ekonomi ve ekonomik kalkınma: Next-11 ülkelerinde su ürünleri yetiştiriciliği ve avcılığının işlevi

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Abstract: Capture fisheries and aquaculture are sectors that provide important benefits and support for the economic development of the countries. Especially for developing countries, these sub-sectors of blue economy are of great importance. The aim of the study is to investigate the impact of capture fisheries and aquaculture sector on economic development in the Next 11 countries for the period from 1990 to 2022. To achieve this goal, we use the novel Method of Moments Quantile Regression (MMQR) approach. The Driscoll-Kraay estimator was also used in the research to test for robustness. The influence of trade openness, energy use, total labor force participation rate, and value added in forestry, agriculture, and fishery (as a percentage of GDP) on economic development in the Next-11 nations was also examined. According to an examination of the study's MMQR findings, aquaculture and capture fisheries have a positive impact on economic development across all quantiles. The research also found that although trade openness and value added in forestry, agriculture, and fisheries (as a percentage of GDP) decrease economic development, labor force participation rate and energy use in the Next-11 nations increase economic development. Additionally, the robustness test results support these findings. We provide policy suggestions based on our results.

Keywords: Fisheries, aquaculture, economic development, Method of Moments Quantile Regression, blue economy.

Öz: Su ürünleri avcılığı ve yetiştiriciliği, ülkelerin ekonomik kalkınmasına önemli fayda ve destek sağlayan sektörlerdir. Özellikle gelişmekte olan ülkeler için mavi ekonominin bu alt sektörleri büyük önem taşımaktadır. Çalışmanın amacı, 1990-2022 dönemi için Next-11 ülkelerinde su ürünleri avcılığı ve yetiştiriciliği sektörlerinin ekonomik kalkınma üzerindeki etkisini araştırmaktır ve bu amaca ulaşmak için yeni bir yöntem olarak Momentler Kantil Regresyon Yöntemi (MMQR) çalışma kapsamında kullanılmaktadır. Çalışmada ayrıca sağlamlık sınaması için Driscoll-Kraay tahmincisine başvurulmuştur. Next-11 ülkelerinde ayrıca ticari açıklığın, enerji kullanımının, toplam işgücü katılım oranının ve ormancılık, tarım ve balıkçılık sektörlerinde oluşan katma değer GSYİH içindeki % payının da ekonomik kalkınma üzerindeki etkisi incelenmiştir. MMQR analizinden elde edilen sonuçlar, su ürünleri avcılığı ve su ürünleri yetiştiriciliği sektörlerinin Next-11 ülkelerinde tüm kantil düzeylerinde ekonomik kalkınma üzerinde olumlu bir etkiye sahip olduğunu göstermektedir. Araştırmada ayrıca, ticari açıklığın ve ormancılık, tarım ve balıkçılık sektörlerinde oluşan katma değer GSYİH içindeki % payının Next 11 ülkelerinde ekonomik kalkınmayı azalttığı, toplam işgücüne katılım oranı ile enerji kullanımının ise ekonomik kalkınmayı artırdığı tespit edilmiştir. Bununla beraber MMQR yöntemi ile elde edilen tüm bu bulgular, sağlamlık testi ile ulaşılan bulgularla da desteklenmektedir. Çalışmada elde edilen bulgular doğrultusunda politika önerileri ortaya konulmuştur.

Anahtar kelimeler: Balıkçılık, akuakültür, ekonomik kalkınma, Momentler Kantil Regresyon Yöntemi, mavi ekonomi.

INTRODUCTION

Oceans and inland waters (rivers, lakes, reservoirs and seas) provide significant benefits to all humanity in important areas such as aquaculture (hereafter, AC) and fishing, maritime and coastal tourism, mineral extraction and power (Ababouch and Fipi, 2015). It is well known that fish provide a variety of minerals and vitamins essential to human health. Fish is also a significant source of protein (Sun, 2008). Along with all this, animal feed can be produced with fish meal obtained from the fish. Again, the fishing sector contributes to the GDP of countries that are coastal and surrounded by rivers. These countries, which are coastal and surrounded by rivers, can export fish to regional and international markets, thus providing a significant source of foreign exchange. Again, the fish

industry supports local and national government revenues through various taxes, tariffs and licensing fees. For fishing communities living in distant areas, the growth of the fishing industry has also aided in the acquisition of more advanced infrastructural prospects. In this context, physical infrastructure facilities such as roads, schools and hospitals have been improved and social facilities have been rehabilitated both directly and indirectly (Odongkara et al., 2005).

One of the first productive endeavors in the history of humanity is said to be fishing. In this context, archaeological research provides important evidence. Pictures found in the caves show that seafood was important for survival and that

significant amounts of seafood were consumed in those times. Similarly, anthropological research indicates that hunting marine organisms was an important part of the feeding activities of many primitive tribes. Historical sources indicate that fishing was an important production activity in the past as well as today (Amason, 2010). Fish production from the seas and oceans is carried out through capture fisheries (hereafter, CF) and AC sectors. For those who live near streams, the sea, or other bodies of water, the CF and AC sectors are considered a vital source of food and a means of subsistence (Smith et al., 2010). More than a billion people make a living on CF and AC sectors (Asche, 2011).

Through the CF and AC sectors, many countries are able to generate significant revenues through production and trade. These sectors can also create jobs and stimulate economic development (hereafter, ED). Apart from all this, growth in these sectors is also important for food security (Payne, 2000; Phuong and Gopalakrishnan, 2004; Rad and Rad, 2012; Welcomme et al., 2010). Fishing activities contribute to the economic structure by having indirect effects as well as direct effects. Indirect effects of the fishing sector can be observed in many sectors, from the agricultural sector to forestry, from the manufacturing sector to financial services (Dyck and Sumaila, 2010).

The report (FAO, 2025) states that the total amount of aquaculture produced worldwide in 2023 was valued at US\$365.98 billion and hit a record high of 136.16 million tonnes (Mt, live wet weight equivalent). In comparison, the total output from catch fisheries worldwide has been calculated to be 91.74 Mt in 2023 (FAO, 2025, Tacon et al., 2025). It has been noted that there has been a notable reduction in the global fish populations in the last several years. Because of this, it is observed that global fishing activities can no longer produce maximum sustainable yields (Agnew et al., 2009). The reduction in these stocks has encouraged AC in particular. The remarkable increase in the availability of seafood for consumption by individuals is mostly due to AC (Cressey, 2016).

The benefits of CF for humans in the context of nutrition, employment, recreation and socio-ED have been ongoing since 90,000 years ago to the present day. CF, which is the most common way of collecting water resources, has been practiced since prehistoric times (Lackey, 2005). For coastal communities, CF is an important source of food and economic activity. This also makes capture fishing an important determinant of social structure (Ochiewo, 2016). CF sector is of great importance within the scope of food supply. The CF industry may contract if future production growth is not possible and if it is not well managed. This may leave the world with a new problem of food shortages (Garcia and Rosenberg, 2010). Most of the harvest obtained through CF is consumed by humans. The remainder are mostly utilized for other than food or related food purposes, including as the manufacturing of fish oil and meal, which are used as ingredients in animal, pet, and

AC feed (Cashion, 2016).

Although CF offers a very important potential to countries, it also faces significant threats such as illegal fishing activities, overfishing, and the degradation of marine ecosystems (Costello et al., 2020). Efforts are being carried out to prevent these threats within the scope of the concept of blue economy (hereafter, BE) which has been put forward by international organizations in recent years (Çoban and Ölmez, 2017) While global wild fisheries has been almost constant for three decades, potential annual production is actually above the current level. Enhancements in management of fisheries will lead to increased output levels (Costello et al., 2020).

The AC sector offers people around the world to feed on farmed fish (Pauly et al., 2003). Growing AC has a great deal of potential to contribute to feeding the world's population in a sustainable manner (Froehlich et al., 2018). Despite all of this, AC is recognized for concentrating on a small number of species (Sampantamit et al., 2020). For the world food regime, trade is an important factor as well as production. Fish products from CF and AC make up approximately ten percent of the world's agricultural commerce. The worth of the world's fish trade is more than the value of all other foods generated from animals combined (Anderson et al., 2011). Fisheries and aquaculture make up around 1.3% of the world's GDP, according to the FAO (2020). This number might be much higher in developing countries, demonstrating the sectors' significance for economic development (Qase, 2025). The food industry in AC is expanding at the fastest rate in the globe, and the BE agenda aligns nicely with the possibility for significant output increases (European Commission, 2012).

While blue growth (hereafter, BG) is often understood to represent economic expansion in the marine sector, some of its proponents use it to introduce new initiatives in the BE since these initiatives have room to expand. Seen from this angle, the 'conventional sector' of CF seems to have little to no promise for BG. However, some advocates of BG decide to add CF. Some people just consider how CF may be consumed by humans, whereas others consider all forms of CF (Boonstra et al., 2018; Klinger et al., 2017; FAO, 2015; Ehlers, 2016; European Commission, 2012; ECORYS, 2012).

CF and AC are seen as components of the "BE," a concept that maximizes revenue from the wise use of marine resources while minimizing harm to the environment and bolstering social benefits (Farmery et al., 2021). In order to achieve ecologically sustainable economic growth and enhance social equity and human well-being, the BE has been suggested as an ocean-based equivalent to the Green Economy development plan (Cisneros Montemayor et al., 2019). The CF and AC sector can be crucial for the transition to a BE/ BG because of their interdependence with the aquatic ecosystem and their dependence on them, and because of the potential of the people involved in them to act as not only users of the resource, but as resource managers. Though it would be prudent to go on as normal, in the end, this would not be sustainable, leading

to hazards that would restrict economic progress and impose costs on people and the environment (Ababouch and Fipi, 2015). With its use of maritime areas, the growth of wild CF and AC is inextricably linked to the existence of a lucrative and sustainable BE, which is entirely dependent upon a robust and productive environment (Tsiouvalas et al., 2022).

A few studies in the literature look at how AC and CF affect ED (Esin et al., 2025; Kusdiantoro Fahrudin et al., 2019; Ahammed et al., 2025; Liza et al., 2025; Alharthi and Hanif, 2020). Subsectors of the BE, such as AC and CF, have been shown to support ED in these studies, which were carried out for various nations and country groups.

The nations that make up the Next 11 have a lot of potential since they are located near seas and oceans: Bangladesh, Vietnam, Egypt, Pakistan, Indonesia, South Korea, Iran, Türkiye, Mexico, the Philippines and Nigeria. Examining the impact of the CF and AC sectors on the ED of the Next 11 countries between 1990 and 2022 is the goal of this study. An essential research question will be answered by this study:

What role do the blue economy sectors—such as aquaculture and capture fisheries—play in the Next-11 nations' economic development?

The majority of the Next 11 countries are developing countries. These nations have substantial potential for blue growth. Given the blue economy's potential to create new jobs and its importance in the context of food security, it is necessary to investigate the impact of blue economy sectors on economic development. Motivated by this, the goal is to add to the body of knowledge by examining how the blue economy sectors—capture fisheries and aquaculture—affect economic development in the Next-11 nations over an extensive period of time, from 1990 to 2022. Mentioning the study's further contributions to the literature is also crucial. Regarding the study's another contributions, it may be said that one of its most significant undertakings was the use of a novel methodology. The present research employs a novel approach called "the Method of Moments Quantile Regression" (MMQR), which was developed by Machado and Silva (2019). The study's findings also address a gap in the literature: very few studies have looked at how the BE's subsectors affect ED. Therefore, the goal is to add to the body of knowledge by examining how the sectors of the blue economy affect the economic development of the Next-11 nations. In the end, a comprehensive analysis was made possible by the wide range of methodologies used in the study. By analyzing the effects of variables like trade openness, energy use, the total labor force participation rate, and value added in forestry, agriculture, and fisheries (as a percentage of GDP), along with capture fisheries and aquaculture, on economic development, the study seeks to provide a thorough analysis. With all these contributions, it is expected that the study would fill an important gap in the literature.

MATERIALS AND METHODS

Data set and model

In the study, data from 1990-2022 belonging to Bangladesh, Vietnam, Egypt, Pakistan, Indonesia, South Korea, Iran, Türkiye, Mexico, the Philippines and Nigeria, known as the Next 11 countries, were used and MMQR (Method of Moments Quantile Regression) approach was performed. The availability of data from all the Next 11 countries for the relevant period has been effective in choosing this period as the data range. GDP per capita (constant 2015 US \$) was used as a dependent variable. As independent variables, CF and AC output were measured in metric tons each. In addition to these variables, trade (% of GDP), forestry, agriculture, and fisheries, value added (% of GDP), energy use, and total labor force participation rate were used as control variables. Data on all these variables was taken from the World Bank (2023).

The model laid out within the scope of the study is as follows:

$$\text{LNGDP}_{it} = \alpha_0 + \alpha_1 \text{LNCFP}_{i,t} + \alpha_2 \text{LNAP}_{i,t} + \alpha_3 \text{LNTRD}_{i,t} + \alpha_4 \text{LNFAF}_{i,t} + \alpha_5 \text{LNEU}_{i,t} + \alpha_6 \text{LNTLF}_{i,t} + \varepsilon_{i,t}$$

Every variable has been transformed into its natural logarithmic form. As for the abbreviations expressed in econometric form, GDP shows GDP per capita; CFP shows capture fisheries production; AP shows AC production; TRD shows trade openness rate; FAF shows forestry, agriculture, and fisheries, value added (% of GDP); EU shows energy use; and TLF shows total labor force participation rate.

Variables in the study, code of these variables, description of data and the source of data are seen in Table 1.

Table 1. Details of variables

Code of Variables	Description of Data	Source of Data
LNGDP	GDP per capita (constant 2015 US\$)	World Bank (2023)
LNCFP	CF production (metric ton)	World Bank (2023)
LNAP	AC production (metric ton)	World Bank (2023)
LNTRD	Trade (% of GDP)	World Bank (2023)
LNFAF	Forestry, agriculture, and fisheries, value added (% of GDP)	World Bank (2023)
LNEU	Usage of energy (kg of oil equivalent per person)	World Bank (2023)
LNTLF	Total labor force participation rate (% of the population aged 15–64) (based on an ILO estimate)	World Bank (2023)

Method

In the study, the analysis was carried out with the MMQR method, which is a novel method. This novel methodology was developed by Machado and Silva (2019). The main benefit of this MMQR model is that, unlike many other approaches, it is capable of capturing the heterogeneity and distributional differences between the reaction and explanatory variables at various quantiles (Jahanger et al., 2023). This makes it simpler

to see the actual variations in a dependent variable throughout a variety of phases, which linear approaches are unable to capture (Guan et al., 2023). When the model has individual effects and endogenous explanatory variables, this method is also helpful (Adebayo et al., 2022). This approach is believed to be helpful in managing any outliers that might distort the distribution of the data overall (Hieu and Mai, 2023). This approach is far more effective in estimating and drawing conclusions from the estimates than the conventional approaches. Traditional panel data estimation methods use the average of the whole panel data to evaluate the quantiles of the regression, ignoring the variability of

individual cross-sections. If the heterogeneity is considered, though, the estimation processes may yield more accurate insights. This problem is handled by the MMQR approach, which is by far the best estimating methodology when compared to traditional methods (Yang et al., 2022).

RESULTS

The selected variables' descriptive statistics are shown in Table 2. Table 2 lists the total number of observations, standard deviations, mean values, lowest and maximum values.

Table 2. Descriptive statistics

Variable	Number of Observations	Mean	Std. Dev.	Min.	Max.	Skewness	Kurtosis	Jarque Bera
LNGDP	363	8.061262	0.9720066	6.160036	10.48596	0.4876554	2.66728	13.51***
LNCFP	363	13.83301	0.851284	12.38174	15.79055	0.2937765	2.101805	37.95***
LNAP	363	12.78168	1.732205	8.662505	16.59546	-0.165028	2.420538	10.07***
LNTRD	363	3.914016	0.4418895	2.794362	5.228261	0.5186029	3.326462	14.57***
LNFAF	363	2.442281	0.7351845	0.3704223	3.701034	-0.9456646	3.139352	33.51***
LNEU	363	6.627154	0.9162407	4.704713	8.618265	0.2916867	2.454874	11.55***
LNTLF	363	4.111765	0.1856939	3.754527	4.445705	0.1139386	2.142262	29.29***

Note: The symbols *, **, and *** denote significance at 10%, 5%, and 1%, respectively.

Table 2 shows that the number of observations for each variable is 363. Once again, Table 2 displays the mean, standard deviation, minimum, and maximum values of the variables that were part of the research. The results of the variables' skewness, kurtosis, and Jarque-Bera tests are also shown in Table 3. All of the chosen series—aside from LNAP and LNFAF—have a positive skew. According to the Kurtosis findings, the variables LNGDP, LNCFP, LNAP, LNEU, and LNTLF have less than normal distributions, but LNTRD and LNFAF have bigger than normal distributions. None of the variables have a normal distribution, according to the Jarque-Bera test findings. As a result, for every series, the null hypothesis that the series has a normal distribution has been rejected. The quantile-based framework is the best strategy for addressing the non-linearity in the data series, according to the Jarque-Bera test statistics (Radulescu et al., 2024).

A graphical tool used in statistics to determine if a dataset follows a certain theoretical distribution, such a normal distribution, is a Q-Q (Quantile-Quantile) plot (Majeed et al., 2023). Q-Q plots, which are used to assess whether or not the data series follow a normal distribution, are seen in Figure 1.

The multicollinearity problem, known as the relationship of two or more independent variables to each other, may be important according to its extent. The fact that the multicollinearity problem is minimal indicates that this problem can sometimes cause large problems, while the fact that this problem is at a medium or high level indicates that analysis should not be performed without overcoming this problem (Daoud, 2017). Although there are a number of methods for testing multicollinearity, the most common one

among these methods is the VIF test. In this study, the VIF test was expressed to investigate multicollinearity. The VIF test results are shown in Table 3.

Table 3. VIF test

Variables	VIF	1/VIF
LNEU	4.62	0.216340
LNFAF	4.11	0.243426
LNCFP	2.87	0.348058
LNAP	2.30	0.435324
LNTLF	1.83	0.546939
LNTRD	1.75	0.570911
Mean		2.91

Table 3's VIF test results show that the VIF values are below 10. Furthermore, the 1/VIF values are higher than 0.10. Multicollinearity is not a concern in the model since the VIF values are fewer than 10 and the 1/VIF values are more than 0.10.

Panel data models have a crucial problem with cross-sectional dependence (CD), and if it is not properly handled, the results might become erroneous, ineffective, and unsatisfactory (Ng et al., 2020; Jahanger et al., 2023).

The Next-11 nations have certain commonalities, which is the rationale for evaluating the CD problem. Therefore, these commonalities may impact the result via a common variance, which must be considered when estimating the unit root, cointegration, and regression analysis (Ahmad and Zhao, 2018; Aziz et al., 2021; Yang et al., 2022). The Pesaran (2015) CD test was used in this investigation to check for reliance on the cross-sectional dependency.

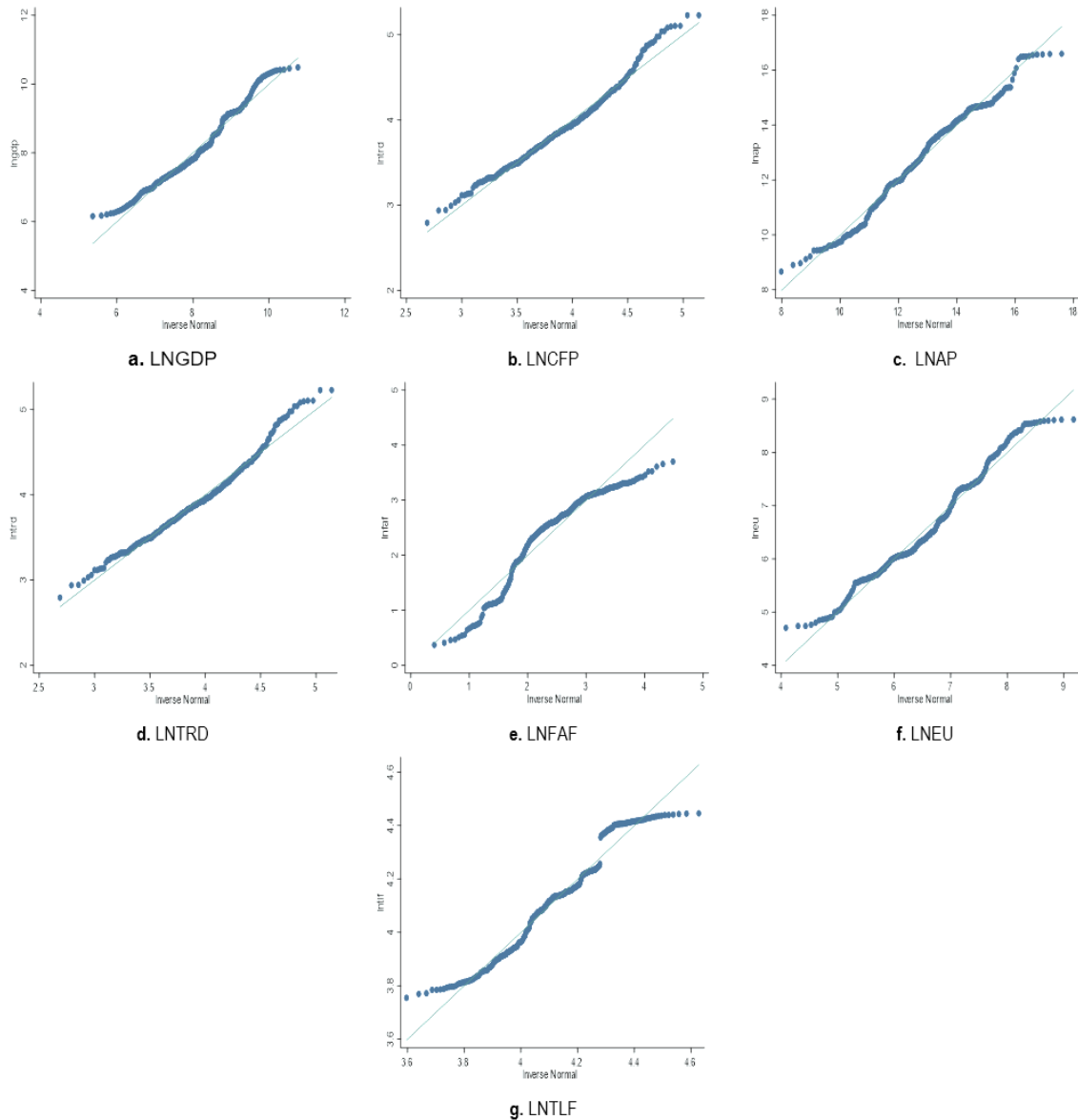


Figure 1. Q-Q Plots, a) LNGDP, b) LNCFP, c) LNAP, d) LNTRD, e) LNFAF f) LNEU, g) LNTLF

Table 4 displays the Pesaran (2015) CD test results. LNGDP, LNCFP, LNAP, LNTRD, LNFAF, and LNEU show cross-sectional dependency at the 1% significance level, according to the Pesaran (2015) CD test findings. Applying second-generation unit root tests that consider cross-sectional dependence comes next (Pesaran, 2007). Consequently, this research uses the Cross-sectionally Augmented Dickey-Fuller (CADF) test. The Cross-sectionally Augmented Dickey-Fuller (CADF) test was developed by Pesaran (2007).

Table 5 displays the results of the CADF test. Every variable series is integrated at first order $I(1)$ at the 0.01% significance level, according to the CADF unit root test. The next stage is to look at the long-term connection between the variables since all of them are stationary in the first difference. To determine the long-term association

between LNGDP, LNCFP, LNAP, LNTRD, LNFAF, LNEU, and LNTLF among the panel nations, we use two cointegration tests. We use two panel cointegration tests: the Pedroni (2004) cointegration test and Kao (1999) cointegration test.

Table 4. Pesaran (2015) CD test

Variables	Statistic	p-value
LNGDP	39.22	0.000***
LNCFP	5.85	0.000***
LNAP	39.41	0.000***
LNTRD	5.04	0.000***
LNFAF	22.99	0.000***
LNEU	26.57	0.000***
LNTLF	0.16	0.873

Note: The symbols *, **, and *** denote significance at 10%, 5%, and 1%, respectively.

Table 5. CADF test

Variables	CADF Level	First Difference
LNGDP	-1.950	-2.464***
LNCFP	-2.067	-3.357***
LNAP	-2.185*	-2.711***
LNTRD	-1.568	-3.157***
LNFAF	-1.645	-3.291***
LNEU	-1.964	-2.863***
LNTLF	-1.660	-2.863***

Note: The symbols *, **, and *** denote significance at 10%, 5%, and 1%, respectively.

The Pedroni (2004) panel cointegration test results are shown in Table 6. The null hypothesis, which states that there is no cointegration relationship among the variables for two of the three sub-tests of the Pedroni test, is rejected. As a result, the variables have a long-term relationship. The Kao (1999) cointegration test was performed in addition to the Pedroni (2004) cointegration test.

Table 6. Results of the Pedroni (2004) cointegration test

Test Statistics	Statistics	Probability
Modified phillips-perron t	2.3833	0.0086***
Phillips-Perron t	-2.0640	0.0195**
Augmented Dickey Fuller t	-0.5704	0.2842

Note: The symbols *, **, and *** denote significance at 10%, 5%, and 1%, respectively.

Table 7 displays the Kao (1999) cointegration test results. The null hypothesis, which asserts that there is no cointegration among the variables, was rejected in each of the test's five subtests, according to an analysis of Kao's (1999) cointegration test results. The results of Kao's (1999) cointegration test, therefore, also support the presence of a long-term link between the variables. The MMQR approach is

Table 8. Estimation findings of MMQR approach

Variables	Location	Scale	Quantiles								
			0.10	0.20	0.30	0.40	0.50	0.60	0.70	0.80	0.90
LNCFP	0.0932404***	-0.0312451**	0.141414***	0.1268432***	0.1151478***	0.1028825***	0.0910658***	0.0818881***	0.0719875***	0.0584244**	0.0403847
LNAP	0.0669559***	-0.0014004	0.069115***	0.068462***	0.0679378***	0.0673881***	0.0668585***	0.0664471***	0.0660034***	0.0653955***	0.064587***
LNTRD	-0.1245868***	-0.0073004	-0.113331***	-0.1167354***	-0.1194681***	-0.1223339***	-0.1250949***	-0.1272392***	-0.1295525***	-0.1327215***	-0.1369365***
LNFAF	-0.4452864***	0.0109667	-0.4621948***	-0.4570806***	-0.4529756***	-0.4486707***	-0.4445232***	-0.4413019***	-0.4378269***	-0.4330664***	-0.4267347***
LNEU	0.5304214***	0.0195721	0.5002453***	0.5093725***	0.5166985***	0.5243816***	0.5317836***	0.5375326***	0.5437343***	0.5522303***	0.5635304***
LNTLF	0.8080071***	0.0792845	0.6857665***	0.7227399***	0.7524171***	0.7835402***	0.8135251***	0.8368137***	0.8619364***	0.8963527***	0.9421284***

Note: The symbols *, **, and *** denote significance at 10%, 5%, and 1%, respectively.

Following the presentation of the MMQR analysis results, a robustness test must be carried out. The Driscoll-Kraay (1998) estimator was used in the study's analysis for robustness testing. By determining the cross-sectional dependencies in the panel data analysis, the DK approach re-estimates the standard errors, guaranteeing the precision and dependability of the computed coefficients (Ince and Ecevit, 2026).

Table 9 displays the outcomes of the Driscoll-Kraay (1998) estimator. These results show that an increase in LNCFP of 1% raises LNGDP by 0.0932404%. A 1% rise in LNAP increases LNGDP by 0.0669559%, according to another finding obtained through the Driscoll-Kraay approach. These

used to estimate the model in the study's subsequent phase.

Table 7. Results of the Kao (1999) cointegration test

Test Statistics	Statistics	P-value
Modified Dickey–Fuller t	-3.1130	0.0009***
Dickey–Fuller t	-2.6749	0.0037***
Augmented Dickey–Fuller t	-2.8617	0.0021***
Unadjusted modified Dickey–Fuller t	-2.5401	0.0055***
Unadjusted Dickey–Fuller t	-2.4681	0.0068***

Note: The symbols *, **, and *** denote significance at 10%, 5%, and 1%, respectively.

The MMQR test results are shown in Table 8. The Next-11 nations' economic development was shown to be positively impacted by capture fisheries across all quantiles, as shown in Table 8. With the exception of the 0.90th quantile, all quantile levels' results are statistically significant. It was determined that there is a positive relationship between the variables across all quantiles based on findings showing the relationship between aquaculture and economic development. Furthermore, all quantile results are statistically significant. The MMQR findings show that the trade openness has a negative impact on economic development across all quantiles in the Next 11 nations. Additionally, this result is statistically significant across all quantiles. For every quantile, a negative relationship between the LNFAF and LNGDP variables was discovered. This result is statistically significant for all quantiles. Energy use has been proven to promote economic growth across all quantiles, according to another finding. Additionally, this result is statistically significant for every quantile. For every quantile, a positive relationship between the LNTLF and LNGDP variables was discovered; it should be noted that this result is statistically significant for every quantile.

results are in line with the MMQR method's findings. Blue economy sectors like aquaculture and capture fisheries have boosted economic development in the Next-11 countries. Additionally, the results are statistically significant. According to the other results, LNGDP is decreased by 0.1245868% for every 1% rise in LNTRD and 0.4452864% for every 1% increase in LNFAF. These results align with the MMQR findings and are statistically significant as well. The research also discovered that a 1% rise in LNEU increases LNGDP by 0.5304214%, while a 1% increase in LNTLF increases LNGDP by 0.8080071%. Additionally, these results are statistically significant. Furthermore, it matches the results of the MMQR.

Table 9. Estimation findings of Driscoll-Kraay Estimator

Variables	Driscoll-Kraay Estimator			
	Coefficient	Driscoll-Kraay Std Errors	t	p-value
LNCFP	0.0932404	.0340325	2.74	0.010***
LNAP	0.0669559	.0096739	6.92	0.000***
LNTRD	-0.1245868	.0314494	-3.96	0.000***
LNFAF	-0.4452864	.0700483	9.98	0.000***
LNEU	0.5304214	.0531696	4.44	0.000***
LNTLF	0.8080071	.1820736	-6.36	0.000***

Note: The symbols *, **, and *** denote significance at 10%, 5%, and 1%, respectively.

DISCUSSION

The blue economy has significant value for emerging nations. The blue economy and its related industries have the potential to propel economic development. The objective of this research is to look at how the Next-11 countries' economic development is affected by blue economy sectors like aquaculture and capture fisheries. The influence of factors including trade openness, energy use, the total labor force participation rate, and value added in forestry, agriculture, and fisheries (as a proportion of GDP) on economic development was also examined. The Next 11 nations' data from 1990 to 2022 are used in the analysis. The data was analyzed using the MMQR technique. Analysis of the study's results showed that, across all quantiles, capture fisheries boost economic development in the Next 11 nations. The finding of this investigation seems to be consistent with a large number of findings from related research published in the literature (Alharthi and Hanif, 2020; Kusdiantoro Fahrudin et al., 2019; Neiland et al., 2016; Rehman et al., 2019). Except for the result for the 0.90th percentile, these results are statistically significant. There is a positive relationship between the variables across all quantiles, according to an analysis of the data showing the connection between aquaculture and economic development. This finding is in line with the vast majority of findings from similar studies in the literature (Alharthi and Hanif, 2020; Rehman et al., 2019; Ahammed et al., 2025; Huyen et al., 2021; Liza et al., 2025). The results are statistically significant. According to other findings obtained using the MMQR method, trade openness and value added in forestry, agriculture, and fisheries (as a percentage of GDP) reduce economic development in all quantiles, while energy use and the total labor force participation rate increase economic development in all quantiles. These results are all statistically significant. Additionally, all of these MMQR results are in line with the robustness test results from Driscoll-Kraay.

CONCLUSION

According to this research, which looks at how blue economy sectors affect economic development in the Next 11 countries, sectors like aquaculture and capture fisheries contribute to economic development. The Next-11 nations should put policies in place to create sustainable fisheries management in light of these results. To maintain the beneficial effects of capture fisheries on economic development, regulations targeted at sustainable fisheries management must be developed since certain fish species are at danger of

overfishing. Additionally, strategies to encourage the use modern production methods should be created, and policies targeted at growing the aquaculture industry in the Next-11 nations should be put into action. The Next-11 nations have to concentrate on blue growth policies and endeavor to create sustainable economic development in these nations. Additionally, it is critical that the socioeconomic development of coastal populations in the Next-11 nations be taken into account while developing these blue growth strategies. In order to help secure food security in the Next-11 countries, efforts should be undertaken to encourage small-scale fishing in these nations. Incentive programs must be put in place right away in order to increase the output of small-scale fisheries in these nations. It is imperative that incentive programs be extended in order to increase the output of small-scale fisheries in these nations.

One of the study's limitations is that a larger range of data could not be examined because of data limitations. Another limitation of the study is that certain variables could not be included in the econometric model due to data constraints in some countries, such as Nigeria. Future study should look at how blue economy sectors affect economic development in African nations, which have not yet received much attention.

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AUTHOR CONTRIBUTIONS

Asım Çoban: Writing—original draft, writing—review & editing, investigation, resources, supervision. Ömer Faruk Biçen: Writing—original draft, writing—review & editing, investigation, data curation. Mustafa Necati Çoban: Writing—original draft, writing—review & editing, supervision, data curation, methodology, software, formal analysis, project administration, investigation, resources, visualization. Fatih Yeter: Writing—original draft, writing—review & editing, investigation, conceptualization. Ayşe Ölmez: Writing—original draft, writing—review & editing, investigation, validation.

CONFLICT OF INTEREST

There are no relevant financial or non-financial interests that the authors need to declare.

ETHICAL APPROVAL

This research does not need clearance from an ethical committee.

DECLARATION OF AI USE

AI-assisted language editing tools (e.g., Grammarly) were used for grammatical and stylistic improvements of the English text.

DATA AVAILABILITY

The datasets used and/or analyzed during the current study are not publicly available due to data use restrictions but are available from the corresponding author on reasonable request.

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