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Do Islamic versus conventional banks progress or regress in productivity level?

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Abstract

This study assesses the differences between Islamic and conventional bank's productivity. Earlier studies on bank productivity focused on conventional banks, but few have been done on Islamic banks. Therefore, the present study attempts to close the gap in the literature by investigating the productivity of Islamic and conventional banks in the context of the Middle East, Southeast Asia and South Asia regions. The sample is comprised of 385 banks (66 Islamic banks and 319 conventional banks) from 18 countries with data observations from 2008 to 2017. Panel data techniques with DEA-based MPI will be employed to investigate the impact of selected important factor and bank productivity as indicated by total factor productivity changes (TFPCH). Based on the results, Islamic banks are more productive than conventional banks and the results from *t* test are further confirmed by the results from nonparametric tests. These results are attributed to the progress in EFFCH. However, the mean difference between Islamic and conventional banks TFPCH is not statistically significant in all regions. The main benefit is that this work will hopefully provide additional insight and complement the existing studies on bank productivity of Islamic and conventional banks that are important to the banks, regulations, investors and researchers.

Keywords: Total factor productivity change, Malmquist productivity index, Islamic bank, Conventional banks, Middle East, Southeast Asia, South Asia

JEL Classifications: G21, G28

Introduction

Nowadays, the banking industry continues to grow, at least, until another form of banking becomes available, and Islamic banking begins to gain further attention from Islamic and contemporary economists. Islamic finance is important for Muslims who require financial instruments that follow the Islamic legal code called *Shariah* [50]. Moreover, Islamic banks also have the potential to reduce risks endemic to financial transaction which have implication for economic growth [28].

The inception of Islamic banking began more than 30 years ago, and since this time, the aggregate of banks

offering financial services has significantly grown to over 300 banks nowadays in over 75 countries from only one bank in 1975 [24]. Islamic finance is important for Muslims who require financial instruments that follow the Islamic legal code called *Shariah* [50]. Also, as the need for *Shariah*-compliant financial products and services increases, this sector will continue to rapidly grow in Muslim and non-Muslim market segments [13]. Referencing Table 1, it can be seen that within 9 years between 2009 and 2018, the total *Shariah* compliance assets significantly increased. Iran, Saudi Arabia, Malaysia, the United Arab Emirates and Kuwait remain the primary markets for Islamic financial banking based on total *Shariah* compliance assets. Notably, South Asia and Southeast Asia countries, including the Middle East, dominate Islamic financing from a global perspective [4]. Therefore, taking these findings into

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Table 1 Ten leading countries for Shariah-compliant assets. Source: The Asian Banker [59] and IFDI (2019)

2009			2018		
Rank	Country	Shariah-Compliant Assets (\$ billion)	Rank	Country	Shariah-Compliant Assets (\$ billion)
1	Iran	236.43	1	Iran	488
2	United Arab Emirates	67.31	2	Saudi Arabia	390
3	Malaysia	56.22	3	Malaysia	214
4	Saudi Arabia	55.01	4	United Arab Emirates	194
5	Kuwait	55.01	5	Kuwait	100
6	Bahrain	38.58	6	Qatar	97
7	Qatar	19.01	7	Turkey	39
8	Turkey	13.76	8	Bahrain	35
9	Sudan	6.12	9	Bangladesh	36
10	Bangladesh	7.5	10	Indonesia	28

account, it is important to comprehend the productivity and nature of both conventional and Islamic banking in these countries.

In principle, Islamic financial system abolishes interest, gambling, speculation, excessive uncertainty (gharar) and illegitimate transactions that are related to alcohol, tobacco, pornography and other activities which considered to be detrimental to the society [27]. Theoretically, Islamic banks which offer Shariah-compliant products can be separable from conventional bank associated with the difference in complexity, agency cost, level of maturity and development [11]. Further differences are evident concerning risk-taking, the price of money, income, the size of banks and so forth [26].

As Islamic banking is viewed as one of the rapidly rising markets, the industry is exposed to credit risk by lengthening borrowing via *Murabaha* and *Ijarah* in generating greater profitability. Furthermore, with non-standard financial agreements and varied methods of funding and intricacies related to managing risk brought about via the introduction of Shariah present further issues regarding Islamic banking to remain stabilised [36]. Islamic banking has the opportunity to utilise profit and loss sharing with respect to liabilities which intensifies credit risk [44]. Moreover, on the balance sheet's liability side, Islamic banks receive deposits founded on profit and loss sharing where they need to pay a profit while investing funds identified on the asset side. Given insufficient investment opportunities, Islamic banks typically have excessive liquidity assets, so liquidity risk is minimal [29]. While Islamic and conventional banks can be distinguished from each other, both banks have a same objective that prioritise the profitability. Simply put, productivity in generating profitability is an important area for both Islamic and conventional banks.

In light of the importance of profitability to Islamic and conventional banks, several studies have focused on analysing the efficiency of Islamic finance as a means of measuring the performance of banks [39, 50, 51, 57, 58, 62], whereas scant research has been undertaken to examine the productivity levels of the Islamic and conventional banking sector as intermediaries [39]. Therefore, this paper will provide a better understanding and contribute to the literature relevant to productivity in banking sector.

Siddiqi [53] mentions that Islamic economic and finance theories are still underdeveloped. Therefore, this study proposes using real-life data to validate the foundational theories in terms of productivity perspective. Productivity is one of the crucial dimensions to measure the firm's performance [15]. The financial performance is a broad concept that takes into consideration of the productivity, profitability and growth. Profitability is the overall efficiency of the company that shows the ability of the firm to earn a profit. One of the primary goals of banks is to look after the interests of shareholders by maximising their return on investment and optimising profits. In order for the bank to get high profit, they need to be productive. However, some of study found that the lower level of productivity can lead to lower profitability of the banks [15]. So, this becomes the main issue. The productivity becomes the main issue because productivity can lead to the lower and higher profitability. So, banks cannot ignore the productivity if they want to increase their profit.

In the context of Islamic banks, similar goals are afforded as to those of conventional banks in maximising profits, although as mentioned previously, both differ as intermediaries. Conventional banks operate on an interest-based principle, whereas Islamic banks adhere to the

interest-free principle by replacing it with profit and loss sharing in performing day-to-day operations founded on Shariah rules and principles [7], although to maximise profits, banks need to be productive, which is directly linked to their efficiency by converting inputs such as capital, raw materials and labour into outputs. Accordingly, as the throughput of outputs increases more rapidly compared to inputs, the efficiency and productivity of banks will also increase. In this situation, productivity determines the level of efficiency in utilising resources (input) to produce outputs.

Globally, the market share of Islamic banking in the financial industry remains low, but in many regions is quickly growing, mainly in the Asian and Middle East regions [33]. According to Houben [31], Southeast Asia, with its ever-increasing Muslim population, receives minimal attention globally. This fact is also supported by Kamarudin et al. [39]. As Islamic finance is increasingly becoming an institutionalised part of the global capital market, it has the distinct potential to rapidly expand and contribute to economic growth [32]. In this context, it is therefore important for Islamic banks to remain productive, in order to remain competitive and contribute towards economic growth. Therefore, instead of just focusing on conventional banks, it is more justifiable if not beneficial, to compare the productivity of Islamic banks.

The lack of comprehensive study on productivity of Islamic and conventional banks inspires this study to investigate the level of productivity amid Islamic and conventional banking sector in the three regions, consisting of South Asia (SA), Southeast Asia (SEA) and Middle East (ME) regions. This study also serves as a continuation to the ongoing debate on whether Islamic banks are more productive relative to conventional banks and vice versa. Hence, this paper focuses on the question whether the productivity of Islamic bank differs from the productivity of conventional bank?

This paper begins with a brief review of related studies and followed by data and methodology, empirical results and conclusion.

Literature review

The role of banking sector as a form of financial intermediation which is part of the financial institution cannot be readily ignored, given it leads to stable economic growth and development. Endogenous growth theory argues that economic growth is primarily result of internal forces rather than external ones through the channel of productivity can be tied directly to faster innovation and more investments in human capital [61]. Business companies wish to be in a position to regulate their spending and overheads, in order to generate greater profits for

stakeholders, similarly, conventional and Islamic banks also seek to improve their productivity given the contribution they make towards economic growth and sustainability. The Islamic banking system plays a similar role but differs slightly compared to the conventional banking system. Islamic banks are considered as a replacement or an alternate option in the provision of banking product and services in accordance with Islamic (Shariah) principles.

Nevertheless, the theory in this area fails to make any obvious prediction as to whether Islamic banks should be more efficient or productive compared to conventional banks [11]. Although productivity is extremely important in Islamic banking in order to gain high profitability, given financing decisions are based on the productivity of the investment in the selected project. Moreover, the Shariah Advisory Committee plays a key part in this respect Islamic banks, in affirming the behaviour stakeholders' in following the principles that govern Islamic law. Moreover, the Shariah Advisory Committee is ultimately responsible for minimising information asymmetries and agency costs within an Islamic bank. According to Jensen and Meckling [34], the existence of conflicts of interest between the principal and agent can influence the performance of organisations.

In this regard, information asymmetry and agency conflicts should be less in Islamic banks compared to their conventional counterparts [48, 60]. Therefore, with the intervention of the Shariah Advisory Committee in monitoring the operations of Islamic banks, conflicts between the principal and agent(s) can be prevented and reduce the agency costs. Ang et al. [6] found that external monitoring produces lower agency costs and thereby increasing the efficiency of banks, leading to high productivity. On the other hand, the opposite may occur given the effect of various productivity determinants are distinctly different in Islamic banks compared to conventional banks such the complexity, level of maturity and development.

Kopleman [43] defines productivity as the relationship between the amount of one or more physical output/s to the associated physical inputs used in production. In other words, they asserted that total production (output) is influenced by the amount of capital invested and the amount of involved labour. Productivity can also be broken down into smaller segments according to Fare et al. [23], based on changes in efficiency or fluctuations in order to compensate for lost ground and also through innovative technological changes, assuming that the outputs are equal equivalent to outputs, and the growth index total factor productivity captures the advancements or changes in technology. Therefore, total factor productivity can be considered equivalent to changes in

technology (from a technical perspective) which can be gauged as a shift in performance, which can accordingly be adjusted by altering the chosen input. Fundamentally, higher productivity will lead to higher bank profitability [39, 54, 56].

In other words, in the context of the banking industry, when the productivity level is increased, the additional output can be therefore be produced from the given amount of input. Accordingly, Cobb–Douglas Production Functions theory will be employed in this study to examine the productivity level between the Islamic and conventional banking sector in the ME, SEA and SA regions. A few studies have found that Islamic banks are significantly more productive compared to conventional banks, other studies have revealed contrary findings and a few studies have revealed that they are similar (no variation) regarding productivity.

More recently, Saleh et al. [52] reveal that the Gulf Cooperation Council (GCC) banks encountered a decline in productivity after the global financial crisis of 2008–2009. They also indicate that the disparity in inefficiency between Islamic and conventional banks has narrowed substantially and that Islamic banks have been able to cross and reduce the gap with conventional banks over the study period from 2005 to 2014. This finding is also supported by Alsharif et al. [5] in which six GCC countries were studied between 2005 and 2015. The findings suggest that Islamic banks are less productive than conventional banks. Moreover, the findings suggest that the Basel III agreement hampered the productivity of the GCC banks, and this detrimental impact is greater for Islamic banks.

Another study by Alexakis et al. [3] reported that in 2008/09, both Islamic and conventional banks experienced a decline in productivity with conventional banks more negatively impacted. The global financial crisis can possibly be connected with this decline in the Gulf Cooperation Council (GCC) banking sector, which was supported by Maredza and Ikhida [45]. The results of this study indicated that there were differences in technological and technical efficiency in Islamic banks, which may be historically related to these banks in the GCC. While this sector is still developing, there are a number of mature Islamic banks in the GCC, although this mix is possibly varied given the assortment of financial products, bank status, clients and innovation.

On the other hand, Rodoni et al. [49] undertook a comparison of the productivity and efficiency of the Islamic banking sector between 2009 and 2013 in Pakistan, Indonesia and Malaysia; the data included 31 banks across these three countries. Using the Malmquist productivity index (MPI) to gauge productivity and Data Envelopment Analysis (DEA) in gauging efficiency the findings

from the study found that the efficiency of the sector in Malaysia was far better compared to Indonesia while in Pakistan, the rate of efficiency was near to 100% during this period.

Kamarudin et al. [39] compared the productivity of 29 Islamic banks in SEA (Malaysia, Indonesia and Brunei) between 2006 and 2014. Using nonparametric DEA-based MPI methods, the researchers approximated the total factor productivity of the banks found that statistically, there was nil variance between the productivity and efficiency and productivity of local and international managed banks given the similar technologies and population.

In another study by Doumpos et al. [21], they investigated the financial robustness of 52 Islamic windows, 347 conventional banks and 101 Islamic banks between 2000 and 2011 by considering 57 member countries of the Organisation of Islamic Cooperation (OIC). The findings from the study indicated that each bank varied concerning the employment of financial ratios, although no difference was evident from a statistical viewpoint for overarching financial robustness between the banks.

Nevertheless, Mobarek and Kalanov [46] examined performance comparatively between Islamic and conventional banks 18 Organisation of Islamic Conference (OIC) countries regarding the pre-global financial crisis period and the actual global financial crisis period between 2004 and 2006 and between 2007 and 2009, respectively. The research was founded on the cross-sectional data of 307 conventional banks and 101 Islamic banks employing DEA and stochastic frontier analysis (SFA) methods in measuring efficiency. The study indicated that the efficiency of conventional banks was higher compared to Islamic banks between 2006 and 2009, which may be due to the mean value of the efficiency score being larger in conventional banks making such a comparison non-equivalent.

Kamarudin et al. [38] investigated the profit, revenue and costs efficiency of 74 banking institutions (47 conventional banks and 27 Islamic banks) in the GCC region between 2007 and 2011. Here, the efficiency level was gauged using the DEA technique by employing the inter-mediation method. The researchers found that conventional banks displayed higher levels of efficiency based on three determinants: revenue, profit and cost. Moreover, they suggest that the primary determinant with respect to the profit efficiency level was the efficiency level associated with revenue.

Therefore, in summary, most studies documented varied and mixed findings on the level of efficiency level amid Islamic banks and conventional banks globally, whereas on the other hand, little has been undertaken, if anything, to explore Islamic and conventional banks

productivity levels. Also, there is less evidence to suggest the productivity level of both types of banks has been conducted in the Asian region, given the strong presence of Islamic banks [39]. Hence based on this deficiency, this research aims to offer evidence empirically mainly on the level of productivity in the Islamic and conventional banks sector.

Methods

Sources of data

The dataset used in this study consisted of Islamic banks and conventional banks from Middle East (ME), South-east Asia (SEA) and South Asia (SA) countries between the period 2008 and 2017 given these three regions are representative of Islamic banking and finance globally [42]. The global financial crisis that occurred between 2008 and 2009 is also taken into consideration in this study, which applies a ‘dummy’ variable to represent this period to avoid any possible biasedness. The sample size of the study consisted of Islamic and conventional banks from 18 countries, (11 from the ME, 4 from SEA and 3 from SA).

The data source employed in this study comprised of information relating to Islamic and conventional banks collected from within the ME, SEA and SA regions between 2008 and 2017. All data were obtained from the Fitch Connect database produced by Fitch Solutions.

Fitch connect is an online repository of data that comprises financial reports, accounting ratios and credit ratings of more 30,000 banks globally including the Islamic and conventional banking sectors. To compare the chosen banking institutions across the three regions (ME, SEA and SA), the currency will be depicted in US dollars.

In total, 385 banks (66 Islamic banks and 319 conventional banks) from 18 countries with dual-banking system are selected in this study as represented in Table 2. Also, to maintain homogeneity, all investment banks, insurance companies and finance companies were excluded in this study. The data of Islamic banks are obtained on the basis of an Islamic subsidiary. The country income level is taken from World Bank Database.

DEA-based Malmquist Productivity Index (MPI)

The data envelopment analysis (DEA) technique was developed by Charnes et al. [18] in which they proposed that the greater the output generated from inputs, the greater the efficiency level associated with the production process. Emrouznejad and Yang [22] affirmed that there had been exponential growth in DEA-related studies, especially given the work of Charnes et al. [18]. Therefore, this confirms that DEA has been acknowledged as a contemporary tool in measuring performance in the diversified fields of management science.

Table 2 Sample data. *Source:* Fitch Connect database

No.	Country	Income group	Region	No. of Islamic bank	No. of conventional bank
1	Bahrain	High	Middle East	8	12
2	Egypt	Lower Middle	Middle East	1	23
3	Iran	Upper Middle	Middle East	1	8
4	Iraq	Upper Middle	Middle East	1	3
5	Jordan	Upper Middle	Middle East	2	11
6	Kuwait	High	Middle East	1	4
7	Lebanon	Upper Middle	Middle East	2	31
8	Oman	High	Middle East	2	7
9	Qatar	High	Middle East	3	5
10	Saudi Arabia	High	Middle East	3	8
11	UAE	High	Middle East	7	14
12	Brunei	High	South East Asia	1	1
13	Indonesia	Lower Middle	South East Asia	8	92
14	Malaysia	Upper Middle	South East Asia	13	31
15	Singapore	High	South East Asia	1	8
16	Bangladesh	Lower Middle	South Asia	4	37
17	Pakistan	Lower Middle	South Asia	7	23
18	Maldives	Upper Middle	South Asia	1	1
	Total			66	319

In the context of this study, efficiency and productivity in an organisation are interrelated, although efficiency is static given it fails to consider the time taken for production, which is important. Accordingly, this shows that when productivity measures alter or change, the level of efficiency also changes. Therefore, it is vital to measure productivity. Fundamentally, the ratio between inputs and outputs can be used to determine productivity.

MPI is occasionally referred to as Total Factor Productivity (TFP), which can assess any change of efficiency and frontier technology in terms of progress or regress over time. Moreover, MPI has been used in many studies for DEA analysis of efficiency changes in diversified fields of management science across various industries and countries.

Output-based MPI was used to understand and gauge the change in the productivity of banks and also to determine the change in TFP Change (TFPCH) to Technical Change (TECHCH) and Efficiency Change (EFFCH). According to Fare et al. [23], the changes in Scale Efficiency Change (SECH) and Pure Technical Change (PTECH) resulted from changes in EFFCH. Figure 1 illustrates the interactive relationship among the efficiency indices

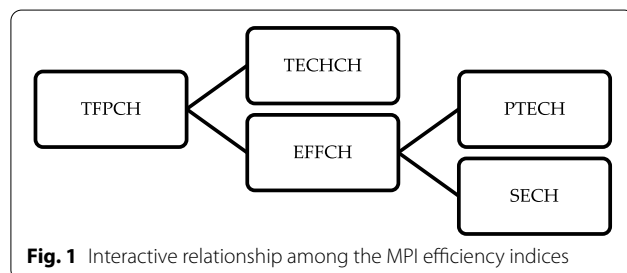
Here, MPI measures the productivity change from period t to $t+1$, which reflects to a reference period technology. Thus, the MPI in relation to technology in period t is:

$$M_0^t = \frac{D_0^t(x^{t+1}, y^{t+1})}{D_0^t(x^t, y^t)}. \quad (1)$$

Corresponding output based MPI concerning technology in period $t+1$ is:

$$M_0^{t+1} = \frac{D_0^{t+1}(x^{t+1}, y^{t+1})}{D_0^{t+1}(x^t, y^t)}. \quad (2)$$

Given it is complicated in choosing among period t and $t+1$ as a benchmark period, an output-based MPI is defined as the geometric mean of Eqs. (1) and (2), [23]:



$$M_0^{t,t+1}(x^{t+1}, y^{t+1}, x^t, y^t) = \left[\left(\frac{D_0^t(x^{t+1}, y^{t+1})}{D_0^t(x^t, y^t)} \right) \times \left(\frac{D_0^{t+1}(x^{t+1}, y^{t+1})}{D_0^{t+1}(x^t, y^t)} \right) \right]^{1/2} \quad (3)$$

Accordingly, this can be broken down into efficiency change (EFFCH^{t,t+1}) as well as technological change (TECHCH^{t,t+1}). As proposed by Fare et al. [23], an equivalent way of writing the MPI index is given below:

$$M_0^{t,t+1}(x^{t+1}, y^{t+1}, x^t, y^t) = \underbrace{\frac{D_0^{t+1}(x^{t+1}, y^{t+1})}{D_0^t(x^t, y^t)}}_{\text{EFFCH}^{t,t+1}} \times \underbrace{\left[\left(\frac{D_0^t(x^{t+1}, y^{t+1})}{D_0^{t+1}(x^{t+1}, y^{t+1})} \right) \times \left(\frac{D_0^t(x^t, y^t)}{D_0^{t+1}(x^t, y^t)} \right) \right]^{1/2}}_{\text{TECHCH}^{t,t+1}} \quad (4)$$

where M represents the level of productivity change as measured by a shift in frontier measured at years' t and $t+1$ in which most of the recent production point (x^{t+1}, y^{t+1}) correspond to the previous production point (x^t, y^t) . However, when $M > 1$ it says that period $(t+1)$ productivity is higher compared to period t productivity. Although when $M < 1$, it says $(t+1)$ productivity is less compared to period t productivity which says that productivity regress and $M = 1.000$ correspond to inaction (no TFP change). Lastly, output distance functions are represented by D 's. The interrelation among the MPI and its two sub-indices can be shown as:

$$M_0^{t,t+1} = \text{Efficiency Change} \times \text{Technical Change} \quad (5)$$

where

$$\text{Efficiency Change} = \frac{D_0^{t+1}(x^{t+1}, y^{t+1})}{D_0^t(x^t, y^t)} \quad (6)$$

Technical Change

$$= \left[\left(\frac{D_0^t(x^{t+1}, y^{t+1})}{D_0^{t+1}(x^{t+1}, y^{t+1})} \right) \times \left(\frac{D_0^t(x^t, y^t)}{D_0^{t+1}(x^t, y^t)} \right) \right]^{1/2} \quad (7)$$

Fare et al. [23] proposed that the index for efficiency change can additionally be broken down into its agreed detailed parts of PTECH ($\Delta \text{PureEff}^{t,t+1}$) measured in relation to the VRS technology and part of SECH ($\Delta \text{Scale}^{t,t+1}$), which captured the change in the variation among the constant returns to scale (CRS) variable and returns to scale (VRS) technologies which can be described as given below:

$$\text{Efficiency Change} = \Delta \text{PureEff}^{t,t+1} \times \Delta \text{Scale}^{t,t+1} \quad (8)$$

where

$$\Delta \text{PureEff}^{t,t+1} = \frac{D_{\text{VRS}}^{t+1}(x_j^{t+1}, y_j^{t+1})}{D_{\text{VRS}}^t(x_j^t, y_j^t)} \quad (9)$$

$$\Delta \text{Scale}^{t,t+1} = \frac{D_{\text{CRS}}^{t+1}(x_j^{t+1}, y_j^{t+1}) / D_{\text{VRS}}^{t+1}(x_j^{t+1}, y_j^{t+1})}{D_{\text{CRS}}^t(x_j^t, y_j^t) / D_{\text{VRS}}^t(x_j^t, y_j^t)} \quad (10)$$

The analysis can also be used to examine the losses or gains relating to the productivity sources to compare the values of both TECHCH and EFFCH. When $\text{EFFCH} > \text{TECHCH}$, it then means that the gains in productivity are primarily from the improvement in efficiency. While when $\text{EFFCH} < \text{TECHCH}$, the gains in productivity are mainly resulting from technological progression.

Therefore, to summarise the analysis of the first stage the TFPCH of banks were determined by employing output-based MPI. Next, VRS technology was used to measure TFPCH ($M_0^{t,t+1}$) relative to efficiency change (EFFCH) and technical change (TECHCH) as given in Eq. (5). Moreover, as proposed by Fare et al. [23], efficiency change (EFFCH) was then broken down further into the element of pure technical change (PTECH) which was determined based on the VRS technology. Accordingly, this was part of the scale efficiency change (SECH) employed to capture the variance among the constant returns to scale (CRS) and variable returns to scale (VRS) technologies as shown in Eq. (8).

The scores representing efficiency were constrained in order to remain amid zero and one and the year 2007 was used as the reference year. The MPI and its constituent parts started with a value of 1.000. Therefore, an efficiency value less (higher) than one for a bank in the following years meant that it was performing below (above) the frontier. Also, the value representing efficiency showed the radial distance from the estimated production frontier to the decision-making unit (DMU) being under consideration.

Specification of banks input and outputs

In order to study productivity, data envelopment analysis (DEA) is employed in this study as a primary tool because it is widely used and still relevant for measuring the productivity given it has been proved to be sustained over time for 40 years with more than a thousand papers published in a year [22].

Accordingly, the intermediation approach was adopted in this study in classifying the input and output of banks, as supported in many studies [12]. This approach has been widely adopted as the initial stage of DEA, given the significant part that banks enact in providing financial intermediation.

In this study, the selection of inputs and outputs was steered by the process as depicted in several studies [3, 19, 39, 55]. All variables employed in the nonparametric DEA were based on the MPI model as part of the initial stage of analysis, as depicted in Table 3.

According to Banker and Datar [10] and Cooper et al. [20], an approximation [assumption] is made in selecting the number of inputs and outputs in that the size of the sample needs to meet this assumption prior to progressing with the measurement of DEA as shown:

$$n \geq \max\{m \times s, 3(m + s)\}$$

where n = number of decision-making unit (DMUs), m = number of inputs, s = number of outputs.

Results

The DEA-based MPI method is employed to examine the objective of this study, which is, to investigate the total factor productivity levels amid Islamic and conventional banking sector in South Asia, Southeast Asia and Middle East regions. The results are then tested using a parametric (t test) and nonparametric (Mann–Whitney [Wilcoxon] and Kruskal–Wallis) test in order to determine the variances in the productivity (y-axis) of Islamic and conventional banks. This test is widely used in prior banking productivity studies [14, 37]. Table 4 presents the summary statistics of data used to construct the productivity frontiers for Islamic and conventional banks. All the variables are measured in US\$m.

Furthermore, Table 5 shows the details results on productivity of Islamic and conventional banks and its decompositions. In addition, the results on Productivity of Islamic and conventional banks using the bank

Table 3 Variables of outputs and inputs. Source: Hassan et al. [30] and Johnes et al. [35]

Variable	Symbol	Variable Name	Definitions
Outputs	y1	Loan	Net loans
	y2	Investment	Total securities
Inputs	x1	Deposits	Total deposits, money market and short-term funding
	x2	Labour	Personnel expenses
	x3	Physical capital	Book value of fixed assets

Table 4 Summary statistics of outputs and input variables in the DEA model (US\$m)

Variables	Output		Input		
	Total loans (y1)	Total Investment (y2)	Deposits (x1)	Labour (x2)	Capital (x3)
<i>Mean</i>					
Conventional banks	7426.564	2787.981	10505.271	106.314	115.847
Islamic banks	4019.186	938.938	5214.859	56.666	76.628
<i>Minimum</i>					
Conventional banks	0.500	0.019	0.800	0.008	0.008
Islamic banks	0.344	0.054	0.979	0.067	0.029
<i>Maximum</i>					
Conventional banks	241,732.006	86,833.757	314,909.471	2113.571	3834.810
Islamic banks	62,276.160	11,346.560	74,287.733	750.373	2095.493
<i>SD</i>					
Conventional banks	21,110.972	7431.583	27,668.656	229.472	274.103
Islamic banks	7274.636	1626.269	8977.123	99.315	177.676

y1 (total of short-term and long-term loans); y2 (total securities); x1 (total deposits, money market and Short-term Funding); x2 (personnel expenses); x3 (fixed assets)

number, specific years, specific regions and different income groups are illustrated in Tables 6 and 7 (see “Appendix”). All the graph levels of productivity in Middle East, Southeast Asia, South Asia and all regions by specific years, specific regions and different income groups are depicted in Figs. 2, 3, 4, 5, 6, 7, 8, 9, 10 and 11.

Discussion

Productivity of Islamic and conventional banks and its decompositions

Table 5 shows the geometric mean scores of the total factor productivity change (TFPCH) and its component, which is the Technical Change (TECHCH) and Efficiency Change (EFFCH) that can be decomposed into Pure Technical Efficiency Change (PTECH) and Scale Efficiency Change (SECH) for all banks (Panel A), conventional banks (Panel B) and IBs (Panel C). This analysis helps to understand the performance of the banks for each year.

Referencing Panel A in Table 5, it can be seen that all banks have, on average, exhibited a lower TFPCH regress of -15.2% (0.848). The results show that all banks exhibited TFPCH regress of -35.9% (0.641) in 2017. During the period of study, the -15.2% (0.848) regress in TFPCH of all banks could be attributed mainly to the -13.4% (0.866) decrease in TECHCH, as the EFFCH seems to have a decrease in the rate of -2.1% (0.979). The decomposition of the EFFCH index consists of PTECH and SECH components indicating that the reason for the decrease in all banks EFFCH was mainly attributed to PTECH rather than SECH. Therefore, considering these findings, all banks are less efficient in the management of

cost control, even though they have been operating at the optimal scale of operations.

Panel B of Table 5 depicts the results for conventional banks. As can be seen in the table, the conventional banks' average exhibited a TFPCH regress of -14.7% (0.853). The results indicate 2017 represents a TFPCH regress of -35.2% (0.648) for conventional banks. The decomposition of the TFPCH index into its TECHCH and EFFCH components reported that the regress in conventional banks' TFPCH was solely attributed to a -12.1% (0.879) decrease in TECHCH, as the EFFCH decreased at a rate of -3.0% (0.970). The decomposition of the EFFCH index into its PTECH and SECH components indicates that the dominant sources that regress in conventional banks' EFFCH were mainly managerially rather than of an operational scale.

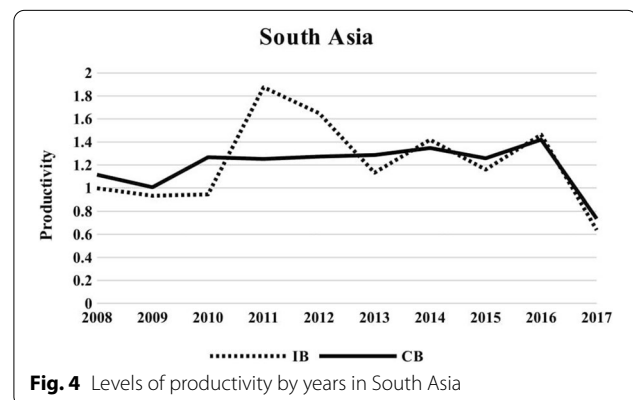
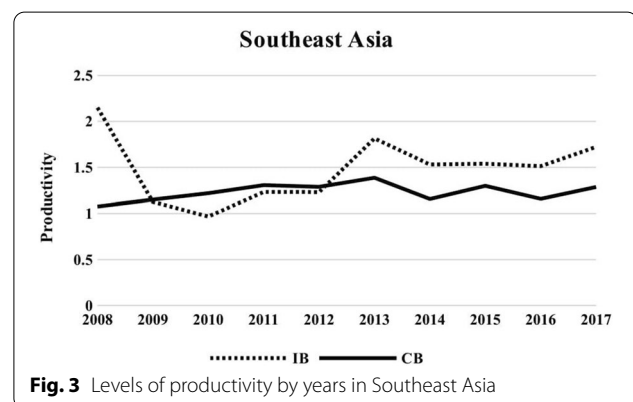
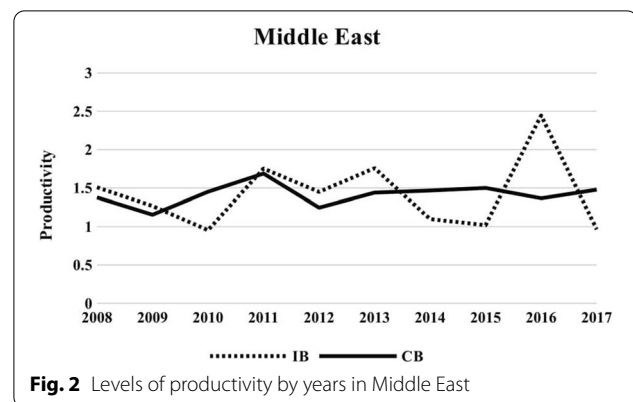
Similarly, Panel C of Table 5 shows the MPI average results of Islamic banks. The empirical findings seem to suggest that the Islamic banks' TFPCH decreased by -11.2% (0.888) higher compared to their conventional banks' counterparts. The results show that for 2017 the TFPCH has a regress of 39.3% (0.607) and progress of 36.1% (1.361) in 2016 for Islamic banks. It can also be seen that within the period of study, the -11.2% (0.888) regress in TFPCH of all banks could be attributed to the -11.0% (0.890) decrease in TECHCH, as the EFFCH seems to have a decline in the rate of -0.2% (0.998). The decomposition of the EFFCH index into its PTECH and SECH components indicates that the decrease in EFFCH was mainly attributed to managerial factors rather than the scale.

Therefore, in summarising the findings, the TFPCH regress in conventional banks and Islamic banks mainly

Table 5 Malmquist productivity index decompositions

Year	INDICES				
	TFPCH	TECHCH	EFFCH	PTECH	SECH
<i>Panel A: All banks</i>					
2007	1.000	1.000	1.000	1.000	1.000
2008	0.973	0.743	1.309	1.239	1.056
2009	0.831	0.818	1.017	0.923	1.102
2010	0.874	0.987	0.885	0.940	0.942
2011	0.995	0.900	1.104	1.034	1.068
2012	0.924	0.977	0.946	0.949	0.997
2013	0.989	0.998	0.992	1.010	0.981
2014	0.841	0.877	0.958	0.874	1.096
2015	0.888	0.935	0.951	0.910	1.044
2016	0.926	0.968	0.957	0.968	1.008
2017	0.641	0.616	1.040	1.056	0.985
Geometric mean	0.848	0.866	0.979	0.957	1.023
<i>Panel B: Conventional banks</i>					
2007	1.000	1.000	1.000	1.000	1.000
2008	0.952	0.732	1.301	1.233	1.054
2009	0.832	0.825	1.009	0.917	1.099
2010	0.905	0.993	0.912	0.960	0.949
2011	1.009	0.901	1.119	1.052	1.064
2012	0.911	0.983	0.927	0.936	0.990
2013	0.981	0.987	0.994	1.019	0.975
2014	0.848	0.875	0.969	0.878	1.104
2015	0.917	0.954	0.961	0.942	1.021
2016	0.859	0.961	0.894	0.889	1.005
2017	0.648	0.597	1.086	1.097	0.989
Geometric mean	0.853	0.879	0.970	0.957	1.014
<i>Panel C: Islamic banks</i>					
2007	1.000	1.000	1.000	1.000	1.000
2008	1.128	0.826	1.364	1.278	1.068
2009	0.829	0.769	1.077	0.964	1.117
2010	0.765	0.866	0.884	0.891	0.992
2011	0.924	0.899	1.025	0.943	1.089
2012	0.995	0.945	1.053	1.022	1.031
2013	1.032	1.049	0.982	0.972	1.012
2014	0.808	0.890	0.908	0.856	1.059
2015	0.763	0.848	0.899	0.773	1.164
2016	1.361	1.005	1.354	1.326	1.021
2017	0.607	0.721	0.843	0.875	0.963
Geometric mean	0.888	0.890	0.998	0.960	1.039

The table presents the geometric mean of the bias-corrected bootstrap total factor productivity change (TFPCH) index and its mutually exhaustive components of technical change (TECHCH) and efficiency change (EFFCH) that is further decomposed into pure technical efficiency change (PTECH) and scale efficiency change (SECH)



result from technological regression (TECHCH). Overall, the empirical findings indicate that the conventional banks and Islamic banks are operating at the optimal scale of operations but are less efficient in the management of cost control.

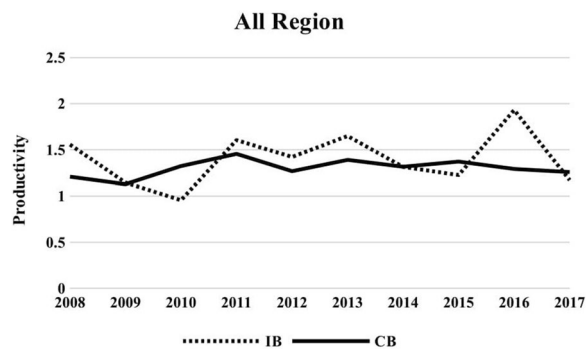


Fig. 5 Levels of productivity by years in all regions

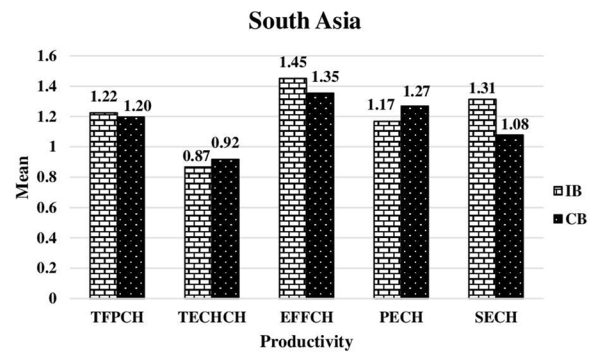


Fig. 8 Levels of productivity in South Asia

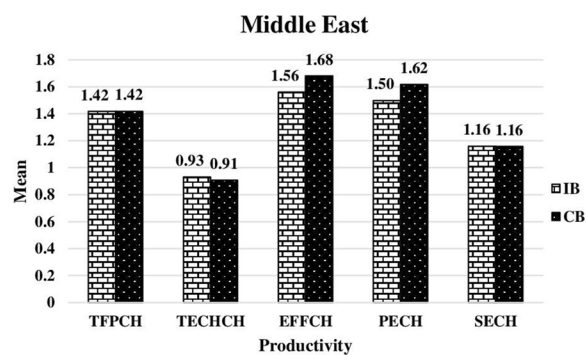


Fig. 6 Levels of productivity in Middle East

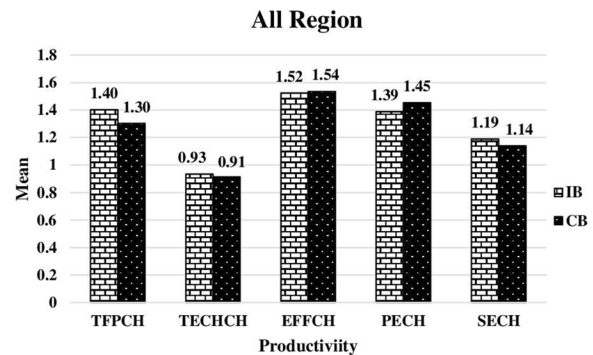


Fig. 9 Levels of productivity in all regions

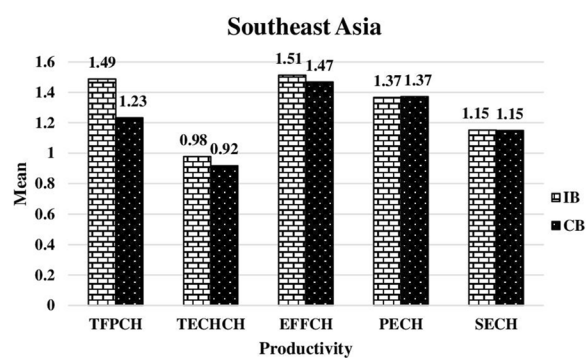


Fig. 7 Levels of productivity in Southeast Asia

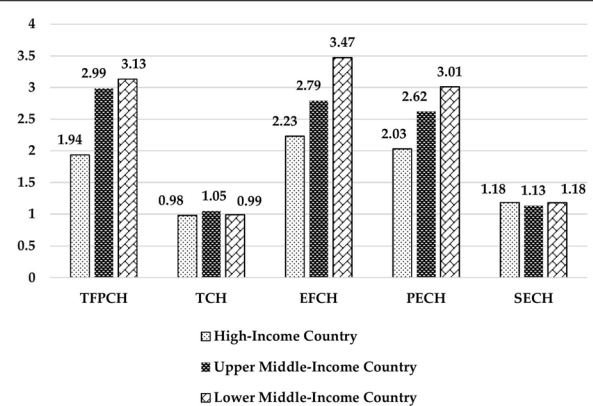
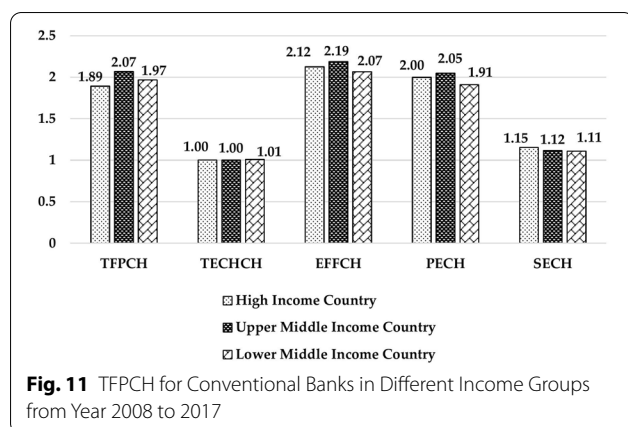


Fig. 10 TFPCH for Islamic Banks in Different Income Groups from Year 2008 to 2017



Productivity of Islamic and conventional banks: an analysis based on the number

In order to control for possible outliers or extreme observations, Table 6 (see “Appendix”) reports the trend in the number and percentage of all banks that experienced productivity progress or regress during the period of study. Referencing Panel A in Table 6, in 2008, only 114 (34.44%) of all banks experienced productivity growth and then increased substantially reaching 198 (55.62%) banks in 2013. However, the number declined to 137 (37.64%) banks in 2017. The result also shows that the number of banks that experienced technological progress increased from 77 (23.26%) banks during 2008 to 196 (55.06%) in 2013 but drastically declined to 81 (22.25%) banks in 2017. Though the number of banks experienced technological regress declined from 228 (68.88%) banks in 2008 to 159 (44.66%) banks in 2013 but before radically increasing to 281 (77.20%) in 2017.

Panel B of Table 6 presents the trend in the number and percentage of conventional banks that experienced productivity progress or regressed between 2008 and 2017. The results suggest that the number of conventional banks experienced productivity progress increased from 92 (31.83%) banks in 2008 to 165 (53.92%) in 2014. However, the number of conventional banks that experienced productivity progress decreased gradually to 114 (37.75%) banks in 2017. Similarly, conventional banks exhibit technological regress decreasing from 200 (69.20%) banks in 2008 to 132 (45.05%) banks in 2013 and before rising to 237 (78.48%) in 2017.

Furthermore, Panel C of Table 6 presents the trend in the number and percentage of Islamic banks that experienced productivity progress or regressed between 2008 and 2017. In 2008, only 22 (52.38%) of Islamic banks

experienced productivity growth, which increased substantially to 36 (57.14%) banks in 2013. However, the number declined to 23 (37.10%) banks in 2017. Likewise, Islamic banks which exhibited a technological progress rise from 10 (23.81%) banks in 2008 to 36 (57.14%) in 2013, reduced to 18 (29.03%) banks in 2017. Further, Islamic banks that exhibited technological regress inclined from 28 (66.67%) in 2008 to 50 (78.13%) in 2014 with the number marginally declining to 44 (70.97%) in 2017.

Productivity of Islamic and conventional banks on specific years

Observing in more detail in the ME region, as reflected in Fig. 2, the productivity indices of Islamic banks progressed in 2016 with 2.445 productivity indices and then regressed in 2010 (0.953). Furthermore, Fig. 3 reveals that productivity indices for both banks in the SEA region follow an inconsistent trend between 2008 and 2017, with productivity indices for Islamic banks peaking in 2008 (2.150) and in 2013 (1.815) staying relatively stable after that. While the Islamic banks reach their peak at 1.875 in 2011 in the SA region as reflected in Fig. 4, before slowly declining in the following year. Islamic banks and conventional banks in SA are rated as having the lowest productivity indices of 0.638 and 0.736 in 2017, whereas the productivity of Islamic banks in the ME and SEA was lower in 2010, at 0.953 and 0.967 productivity indices, respectively. On the other hand, ME conventional banks had the lowest productivity indices at 1.155 in 2009, while conventional banks in SEA had the lowest productivity indices of 1.075 in 2008.

Figure 5 shows the trend related to the productivity of Islamic and conventional banks in all regions between 2008 and 2017. As seen in Fig. 5, the productivity of Islamic banks outperforms conventional banks in most years following the GFC in 2008 and 2009. On average, both Islamic banks and conventional banks have productivity indices above 1.00, indicating that the banks’ have experienced progress in their productivity compared to the prior year. We can also see productivity indices for both banks show the unevenness of the trend between 2008 and 2017, where productivity indices for Islamic banks peaked in 2008 (1.556) and 2011 (1.604), before gradually decreasing from 2013 (1.649) to 2015 (1.226) and reaching a peak in 2016 (1.931). However, the productivity of conventional banks from all regions remains relatively stable in the lowest level range of 1.129 in 2009 to the highest level of 1.455 in 2011.

The reason of this condition is Muslim population is increasing especially in Southeast Asian region which

made substantial popularity in the ethical character and financial stability in Islamic finance product and services as an alternative to conventional financing mechanisms [41]. Moreover, the issuance of global Sukuk in 2012 has been growing in popularity and utilised by both corporate sector and sovereigns for raising fund in South-east Asian country, in particular, Malaysia, followed by Indonesia to promote Shariah-based financial product that enabling these regions to dominate over 70 per cent of the world's Sukuk issuances [41]. Furthermore, South Asia has made investment for their infrastructure development which is one of the major obstacles to the economic growth in year 2016 [8]. Sukuk played an important role in financing the infrastructure projects in South Asia region [41].

To conclude, Fig. 5 shows the overall empirical mean productivity level in all regions indicating that Islamic banks have a higher overall productivity level compared to conventional banks [1.931 (2016) vs. 1.455 (2011)]. This suggests that the Islamic banks are more productive than conventional banks in managing their operations. The level of productivity in Islamic banks in all three regions (ME, SEA and SA) [2.445 (2016), 2.15 (2008), 1.875 (2011)], respectively, is higher than in conventional banks [1.689 (2011), 1.306 (2011), 1.348 (2014)].

Productivity of Islamic and conventional banks in specific regions

Figure 6 represents the productivity component of Islamic and conventional banks in ME countries between 2008 and 2017. The TFPCH are almost similar for both types of banks in the ME; Islamic banks (1.417) and conventional banks (1.418). The TFPCH progress in both banks results from higher EFFCH, which means there is an improvement in the management of the banks. When looking at the decomposition of the EFFCH, both banks have progressed in PECH. Figure 7 shows the productivity component of Islamic and conventional banks in SEA countries between 2008 and 2017. Here, overall, Islamic banks have higher TFPCH (1.488 vs. 1.233), TECHCH (0.977 vs. 0.918), EFFCH (1.513 vs. 1.469) and SECH (1.152 vs. 1.149), while conventional banks have slightly higher PECH (1.372 vs. 1.367) than Islamic banks. The TFPCH in both banks results from the EFFCH. When considering at the decomposition of EFFCH, both banks have progressed in PECH.

Figure 8 confirms the productivity component of Islamic and conventional banks in SA countries between 2008 and 2017. Overall, Islamic banks have higher TFPCH (1.488 vs. 1.233), EFFCH (1.513 vs. 1.469) and

SECH (1.152 vs. 1.149). While conventional banks are higher in TECHCH (0.977 vs. 0.918) and PECH (1.367 vs. 1.372) than Islamic banks. The TFPCH in both Islamic and conventional banks results from EFFCH, same as for other regions. The decomposition of EFFCH shows that both banks have progressed in SECH.

To conclude, Fig. 9 reveals the overall empirical mean productivity level in all regions indicating that Islamic banks have a higher overall productivity level than conventional banks (1.402 vs. 1.302). This suggests that Islamic banks are more productive than conventional banks in handling their respective operations. The level of productivity in Islamic banks in all three regions (ME, SEA and SA) [1.417, 1.488, 1.224] is higher than in conventional banks [1.418, 1.233, 1.196], respectively.

Productivity of Islamic and conventional banks in different income groups

Figure 10 represents TFPCH of Islamic banks in different income groups which are high-income, upper middle-income and lower middle-income countries from year 2008 to 2017. Overall, high-income countries have lowest TFPCH, TECHCH, EFFCH and PECH. While lower middle-income countries have higher TFPCH, EFFCH, PECH and SECH. The upper middle-income countries have slightly higher TECHCH than high-income countries and lower middle-income countries.

The reason of this condition can be related to the increasing number of Islamic banks in the upper middle-income countries and lower middle-income countries such as Malaysia, Indonesia, Pakistan and Bangladesh. High Muslim population with their income levels has a positive effect on the Islamic banking system development [16]. As the number of Islamic banks in these income groups is increasing especially in Malaysia and Indonesia, it creates a highly competitive market among them. According to Abedifar et al. [2], the coexistence of both banks in the market can motivate the banks to be more innovative and increase the efficiency of whole banking system by improving competition.

As we can see from Fig. 10, the TFPCH in all income group countries comes from the EFFCH and this means there is managerial change in the banks. According to Boukhatem and Moussa [16], countries that have adopted mixed legal system based on Common law and Islamic (Shariah) law enable them to be flexible in response to the shift in macroeconomic conditions has contributed to the Islamic banks development.

Figure 11 represents the TFPCH for conventional banks in different income groups from year 2008 to 2017. The upper middle-income countries have slightly higher TFPCH

than high-income group countries and lower middle-income group countries. Most of the changes are come from the EFFCH due to managerial efficiency. This result is in line with the findings of Aluko and Ajayi [1]. They found that lower income countries tend to have more efficient banks than high-income countries. High-income countries have a larger size of banking industry and they face more competitive pressures that can lead to larger agency cost and higher overhead costs, subsequently leading to lower productivity in bank operation [25].

Robustness tests

Table 7 (see “Appendix”) presents the significant difference between the productivity levels of Islamic and conventional banks in specific years and regions derived from the parametric (t test) and nonparametric (Mann–Whitney [Wilcoxon] and Kruskal–Wallis tests) tests. The empirical findings presented in Table 7 suggest that the Islamic banks are marginally more productive than conventional banks in all regions. However, the difference is only statistically significant during 2008 and 2016 (Panels A and I). On the other hand, Panels B, C, G, H and J in the t test suggest that the conventional banks have been relatively more productive than Islamic banks. However, the results are only statistically significant in Panel C. The results from nonparametric tests further confirm the t test results.

Whereas, for the ME region, the results imply that the TFPCH level for Islamic banks is more progressed than in conventional banks in six out of the 10 years under examination. The results are significant at a 5% level in Panel I. However, it can also be observed that conventional banks have higher TFPCH than Islamic banks in the ME region and are statistically significant in Panels C and J. Notwithstanding, Islamic banks have been relatively more progressed in terms of productivity compared to conventional banks in Panels A, E, F, H, I and J in the SEA region. However, only Panels A and I are statistically significant at 10% and 5%, respectively. Nevertheless, Panels D, E, G and I suggest that Islamic banks are more productive than conventional banks in the SA region but only statistically significant at 1% in Panel D. The results also suggest that conventional banks are more progressive than Islamic banks in SA the region in Panels A, B, C, F, H and J. However, the difference is insignificant.

When analysing the results further, the findings also show the component of the MPI, including the TECHCH, EFFCH and decomposition of EFFCH, namely PECH and SECH. Here, the findings show that Islamic are relatively more productive, attributed to higher TECHCH and statistically significant in Panels E, F and J. On the other hand, conventional banks are present more TECHCH and are statistically significant at Panel H. Likewise, the EFFCH in

Islamic banks is higher in Panels D, E, H and I and lower in Panels A, B, C, F, G and J. When we look at the decomposition of EFFCH, Islamic banks are statistically significant to have been progressed in PECH, as reflected in Panels I and J. However, the Islamic banks only experienced SECH progress in Panel H and is statistically significant at the 5% level.

Panel K of Table 7c presents the results based on all years and regions. Based on the results, Islamic banks are more productive than conventional banks with a mean difference of $(1.402 > 1.302 = 0.100)$. The results from the nonparametric tests further confirm the results from the t test. These results are attributed to the progress in TECHCH with a mean difference of (0.019) and statistically significant at 10% levels. The mean difference between Islamic and conventional banks TFPCH is not statistically significant in all regions, though.

The ME, SEA and SA regions also show that Islamic banks are more productive than conventional banks in Panel K of Table 7c. In the ME region, the productivity levels are similar for both types of banks $(1.418 = 1.418)$; though, it is not significant at any level. The difference is only statistically significant in the SEA region with 5% levels. The productivity attributed to higher TECHCH is statistically significant at 1% levels. The results from the parametric t test are further confirmed through the nonparametric Mann–Whitney (Wilcoxon) and Kruskal–Wallis tests.

The results on the means for TFPCH in the Islamic and conventional banks in SA revealed that the TFPCH in the Islamic banks has a higher mean compared to conventional banks $(1.224 > 1.196)$; however, it is not statistically significant. The progression in productivity for Islamic and conventional banks originates from the TECHCH and is significantly different at a 1% level (under the nonparametric test). Overall, the mean difference shows that Islamic banks are more productive than conventional banks in all regions, namely the ME, SEA and SA regions.

Therefore, in summary, it was revealed that Islamic banks tend to be more progressive compared to conventional banks but insignificantly different. This conclusion appears to be consistent with the results of earlier research that Islamic and conventional banks may not completely translate into visible distinctions. Given the benefit that conventional banks enjoy over the Islamic banks, where conventional banks have a very long history and expertise, their main income comes from interest, has very huge capital, spreads broadly and has much more advanced technology, the findings of this study are in favour of the later banking system. Beck et al. [11], for instance, point out that Islamic banks have produced products that emulate conventional banking products, replacing interest rate payments

and discounting with fees and contingent payment arrangement. Likewise, Khan [40] concludes that there are considerable gaps between Islamic banking and finance principles and their real practice, but most of Islamic banking and finance remains technically indistinguishable from conventional banking. In addition, Bourkhis and Nabi [17] also point out that in Islamic bank, only 20% are committed to long-term and risk-sharing funding and argue that Islamic banks emulate conventional banks and diverge from their theoretical business model. The findings of Doumpos et al. [21], Hassan et al. [30], Bader et al. [9] and Mohamad et al. [47] are also in favour of this finding.

Conclusions

This study seeks to add to the literature on banks performance. As the Islamic banking growth is rapidly soaring, it is imperative to study their productivity. This paper aims to analyse total factor productivity change of Islamic and conventional banks covering 18 countries with dual-banking systems in the context of Middle East, Southeast Asia and South Asia regions. By using nonparametric DEA-based MPI method, the data are tested using parametric (t test) and nonparametric (Mann–Whitney and Kruskal–Wallis) tests. Although the theory suggests that there is distinctly difference between Islamic banks compared to conventional banks in the effect of various productivity determinants such the complexity, agency cost, level of maturity and development, empirical estimation suggests that there is no statistically significant difference between the total factor productivity (TFPCH) of the Islamic and conventional banks.

Further examination on the specific years for two banking types shows Islamic banks has exhibited a marginal productivity progress than conventional banks in nearly all the year. Moreover, this study also looked further into specific regions in the context of Middle East, Southeast Asia and South Asia regions. The findings suggest that Islamic banks are more productive than conventional banks in most of the years for all regions. Almost all the progression in productivity of Islamic and conventional banks are solely attributed to the increase in efficiency change (EFFCH). This signifies that both Islamic and conventional banks are efficient in their managerial.

Banks performance has attracted much concern in the banking and finance world. This study has provided important implication for policy makers and business practices. Therefore, determining the overall productivity level of Islamic banks can be a guidance to the policy makers to improve managerial performance, and investors and clients to make other decisions relating to investment and

service quality. The progressive in productivity of the bank is expected to continuously improve the banks' profitability by providing a high-quality service at a minimum cost. One key enable for such commitment would require banking and finance sectors to find the mechanism such as policy matter, workforce trainings and supports to soar up the productivity levels on the institutions. The findings in this study are in line with the existing literature on finance and development.

Furthermore, several limitations are inherent in this study, in which future researchers could explore. First, as this study examined the TFPCH of banks between 2008 and 2017, to achieve a consistent and comparative analysis, relatively little information was collected. As such, there were various aspects of productivity patterns that could not be discussed. Therefore, further research is needed in the future, considering a longer period of time, in order to explore the key determinants of productivity progress in the banking industry.

Secondly, this analysis may be further improved if the ratings of the financial institution are used as one of the factors to assess if higher-rated banks are better than lower-rated banks or vice versa. In addition, the regulatory regime of the various countries may be another consideration. There are well-developed regulations for Islamic banks in some countries, while others do not. Some factors that may be taken into consideration are that Basel III was introduced during the 2007–08 financial crisis, which could have influenced the risk-taking conduct of financial institutions.

Due to its limitations, this study provides the avenue and incentives to seek greater insight for a more deepened understanding and investigation on the productivity levels of two banking types. Future research may consider investigating the different bank production measurements such as production approach and revenue approach as this study is using intermediation approach in determining the specifications of bank's input and output. Besides, the nonparametric frontier analysis can be combined with the parametric stochastic frontier analysis method to attain more robustness empirical evidence. This study, therefore, offers motivations for the institutions to formulate strategic choices to escalate the productivity level of the banks.

Appendix

See Tables 6 and 7.

Table 6 Number and percentage of conventional and Islamic banks with productivity progress and regress

Period	Productivity change			Technological change			Efficiency change			Pure Efficiency change			Scale efficiency change		
	(TFPCH)			(TECHCH)			(EFFCH)			(PECH)			(SECH)		
	Progress	Regress	No Δ	Progress	Regress	No Δ	Progress	Regress	No Δ	Progress	Regress	No Δ	Progress	Regress	No Δ
	# (%)	# (%)	# (%)	# (%)	# (%)	# (%)	# (%)	# (%)	# (%)	# (%)	# (%)	# (%)	# (%)	# (%)	# (%)
<i>Panel A: All banks</i>															
2007–2008	114 (34.44)	89 (26.89)	128 (38.67)	77 (23.26)	228 (68.88)	26 (7.85)	220 (66.47)	88 (26.59)	23 (6.95)	231 (69.79)	77 (23.26)	23 (6.95)	155 (46.83)	131 (39.58)	45 (13.60)
2008–2009	123 (37.27)	143 (43.33)	64 (19.39)	40 (12.12)	283 (85.76)	7 (2.12)	199 (60.30)	119 (36.06)	12 (3.64)	175 (53.03)	137 (41.52)	18 (5.45)	170 (51.52)	107 (32.42)	53 (16.06)
2009–2010	134 (39.30)	155 (45.45)	52 (15.25)	182 (53.37)	157 (46.04)	2 (0.59)	152 (44.57)	181 (53.08)	8 (2.35)	190 (55.72)	134 (39.30)	17 (4.99)	99 (29.03)	202 (59.24)	40 (11.73)
2010–2011	158 (45.93)	146 (42.44)	40 (11.63)	137 (39.83)	206 (59.88)	1 (0.29)	194 (56.40)	136 (39.53)	14 (4.07)	194 (56.40)	126 (36.63)	24 (6.98)	165 (47.97)	128 (37.21)	51 (14.83)
2011–2012	170 (48.71)	145 (41.55)	34 (9.74)	94 (26.93)	254 (72.78)	1 (0.29)	205 (58.74)	132 (37.82)	12 (3.44)	191 (54.73)	135 (38.68)	23 (6.59)	147 (42.12)	154 (44.13)	48 (13.75)
2012–2013	198 (55.62)	148 (41.57)	10 (2.81)	196 (55.06)	159 (44.66)	1 (0.28)	173 (48.60)	170 (47.75)	13 (3.65)	181 (50.84)	156 (43.82)	19 (5.34)	133 (37.36)	181 (50.84)	42 (11.80)
2013–2014	191 (51.62)	175 (47.30)	4 (1.08)	82 (22.16)	288 (77.84)	0 (0.00)	209 (56.49)	148 (40.00)	13 (3.51)	202 (54.59)	147 (39.73)	21 (5.68)	158 (42.70)	161 (43.51)	51 (13.78)
2014–2015	177 (47.97)	187 (50.68)	5 (1.36)	148 (40.11)	220 (59.62)	1 (0.27)	214 (57.99)	145 (39.30)	10 (2.71)	191 (51.76)	157 (42.55)	21 (5.69)	181 (49.05)	139 (37.67)	49 (13.28)
2015–2016	174 (48.20)	180 (49.86)	7 (1.94)	127 (35.18)	232 (64.27)	2 (0.55)	181 (50.14)	165 (45.71)	15 (4.16)	173 (47.92)	166 (45.98)	22 (6.09)	139 (38.50)	173 (47.92)	49 (13.57)
2016–2017	137 (37.64)	225 (61.81)	2 (0.55)	81 (22.25)	281 (77.20)	2 (0.55)	167 (45.88)	192 (52.75)	5 (1.37)	149 (40.93)	205 (56.32)	10 (2.75)	166 (45.60)	191 (52.47)	7 (1.92)
<i>Panel B: Conventional banks</i>															
2007–2008	92 (31.83)	77 (26.64)	120 (41.52)	67 (23.18)	200 (69.20)	22 (7.61)	191 (66.09)	80 (27.68)	18 (6.23)	203 (70.24)	67 (23.18)	19 (6.57)	134 (46.37)	116 (40.14)	39 (13.49)
2008–2009	107 (37.15)	122 (42.36)	59 (20.49)	36 (12.50)	247 (85.76)	5 (1.74)	178 (61.81)	103 (35.76)	7 (2.43)	156 (54.17)	120 (41.67)	12 (4.17)	154 (53.47)	95 (32.99)	39 (13.54)
2009–2010	115 (39.79)	123 (42.56)	51 (17.65)	154 (53.29)	133 (46.02)	2 (0.69)	134 (46.37)	151 (52.25)	4 (1.38)	166 (57.44)	113 (39.10)	10 (3.46)	89 (30.80)	169 (58.48)	31 (10.73)
2010–2011	127 (43.79)	125 (43.10)	38 (13.10)	119 (41.03)	170 (58.62)	1 (0.34)	162 (55.86)	119 (41.03)	9 (3.10)	164 (56.55)	111 (38.28)	15 (5.17)	137 (47.24)	109 (37.59)	44 (15.17)
2011–2012	143 (48.64)	119 (40.48)	32 (10.88)	84 (28.57)	210 (71.43)	0 (0.00)	176 (59.86)	109 (37.07)	9 (3.06)	163 (55.44)	114 (38.78)	17 (5.78)	124 (42.18)	128 (43.54)	42 (14.29)
2012–2013	162 (55.29)	122 (41.64)	9 (3.07)	160 (54.61)	132 (45.05)	1 (0.34)	143 (48.81)	140 (47.78)	10 (3.41)	150 (51.19)	129 (44.03)	14 (4.78)	104 (35.49)	154 (52.56)	35 (11.95)
2013–2014	165 (53.92)	137 (44.77)	4 (1.31)	68 (22.22)	238 (77.78)	0 (0.00)	180 (58.82)	116 (37.91)	10 (3.27)	172 (56.21)	120 (39.22)	14 (4.58)	135 (44.12)	132 (43.14)	39 (12.75)
2014–2015	147 (48.20)	153 (50.16)	5 (1.64)	128 (41.97)	176 (57.70)	1 (0.33)	181 (59.34)	117 (38.36)	7 (2.30)	164 (53.77)	126 (41.31)	15 (4.92)	144 (47.21)	122 (40.00)	39 (12.79)
2015–2016	133 (44.04)	162 (53.64)	7 (2.32)	101 (33.44)	199 (65.89)	2 (0.66)	144 (47.68)	145 (48.01)	13 (4.30)	135 (44.70)	150 (49.67)	17 (5.63)	117 (38.74)	148 (49.01)	37 (12.25)
2016–2017	114 (37.75)	186 (61.59)	2 (0.66)	63 (20.86)	237 (78.48)	2 (0.66)	140 (46.36)	159 (52.65)	3 (0.99)	126 (41.72)	171 (56.62)	5 (1.66)	138 (45.70)	159 (52.65)	5 (1.66)
<i>Panel C: Islamic banks</i>															
2007–2008	22 (52.38)	12 (28.57)	8 (19.05)	10 (23.81)	28 (66.67)	4 (9.52)	29 (69.05)	8 (19.05)	5 (11.90)	28 (66.67)	10 (23.81)	4 (9.52)	21 (50.00)	15 (35.71)	6 (14.29)
2008–2009	16 (38.10)	21 (50.00)	5 (11.90)	4 (9.52)	36 (85.71)	2 (4.76)	21 (50.00)	16 (38.10)	5 (11.90)	19 (45.24)	17 (40.48)	6 (14.29)	16 (38.10)	12 (28.57)	14 (33.33)
2009–2010	19 (36.54)	32 (61.54)	1 (1.92)	28 (53.85)	24 (46.15)	0 (0.00)	18 (34.62)	30 (57.69)	4 (7.69)	24 (46.15)	21 (40.38)	7 (13.46)	10 (19.23)	33 (63.46)	9 (17.31)
2010–2011	31 (57.41)	21 (38.89)	2 (3.70)	18 (33.33)	36 (66.67)	0 (0.00)	32 (59.26)	17 (31.48)	5 (9.26)	30 (55.56)	15 (27.78)	9 (16.67)	28 (51.85)	19 (35.19)	7 (12.96)
2011–2012	27 (49.09)	26 (47.27)	2 (3.64)	10 (18.18)	44 (80.00)	1 (1.82)	29 (52.73)	23 (41.82)	3 (5.45)	28 (50.91)	21 (38.18)	6 (10.91)	23 (41.82)	26 (47.27)	6 (10.91)
2012–2013	36 (57.14)	26 (41.27)	1 (1.59)	36 (57.14)	27 (42.86)	0 (0.00)	30 (47.62)	30 (47.62)	3 (4.76)	31 (49.21)	27 (42.86)	5 (7.94)	29 (46.03)	27 (42.86)	7 (11.11)
2013–2014	26 (40.63)	38 (59.38)	0 (0.00)	14 (21.88)	50 (78.13)	0 (0.00)	29 (45.31)	32 (50.00)	3 (4.69)	30 (46.88)	27 (42.19)	7 (10.94)	23 (35.94)	23 (35.94)	12 (18.75)
2014–2015	30 (46.88)	34 (53.13)	0 (0.00)	20 (31.25)	44 (68.75)	0 (0.00)	33 (51.56)	28 (43.75)	3 (4.69)	27 (42.19)	31 (48.44)	6 (9.38)	37 (57.81)	17 (26.56)	10 (15.63)

Table 6 (continued)

Period	Productivity change				Technological change				Efficiency change				Pure Efficiency change				Scale efficiency change			
	(TFPCH)				(TECHCH)				(EFFCH)				(PECH)				(SECH)			
	Progress	Regress	No Δ		Progress	Regress	No Δ		Progress	Regress	No Δ		Progress	Regress	No Δ		Progress	Regress	No Δ	
	# (%)	# (%)	# (%)		# (%)	# (%)	# (%)		# (%)	# (%)	# (%)		# (%)	# (%)	# (%)		# (%)	# (%)	# (%)	
2015–2016	41 (69.49)	18 (30.51)	0 (0.00)		26 (44.07)	33 (55.93)	0 (0.00)		37 (62.71)	20 (33.90)	2 (3.39)		38 (64.41)	16 (27.12)	5 (8.47)		22 (37.29)	25 (42.37)	12 (20.34)	
2016–2017	23 (37.10)	39 (62.90)	0 (0.00)		18 (29.03)	44 (70.97)	0 (0.00)		27 (43.55)	33 (53.23)	2 (3.23)		23 (37.10)	34 (54.84)	5 (8.06)		28 (45.16)	32 (51.61)	2 (3.23)	

Productivity growth: TFPCH > 1, Productivity Loss < 1, Productivity Stagnation: TFPCH = 1

Table 7 Summary of parametric and nonparametric test on conventional and Islamic banks

Region	No. of Obs.	Parametric test						Nonparametric test					
		t test						Mann–Whitney [Wilcoxon Rank Sum] test					
		t(Prb > t)						z(Prb > z)					
		TFPCH	TECHCH	EFFCH	PECH	SECH		TFPCH	TECHCH	EFFCH	PECH	SECH	
Panel A: year 2007–2008													
ALL	331	–1.449	–1.445	0.140	–0.620	0.232		–1.798 ^c	–1.253	–0.041	–0.419	–0.219	
Mean Diff.		0.347	0.076	(0.030)	0.127	(0.017)		27.580	19.800	(0.650)	(6.630)	3.450	
ME	142	–0.421	0.670	0.105	–0.394	–0.651		–0.038	–0.402	–0.135	–1.279	–0.909	
Mean Diff.		0.132	(0.043)	(0.035)	0.148	0.057		0.360	(3.910)	(1.310)	(12.44)	8.830	
SEA	132	–2.153 ^c	–3.589 ^a	–0.230	–0.378	0.307		–3.644 ^a	–3.072 ^a	–0.301	–0.416	–0.350	
Mean Diff.		1.076	0.384	0.088	0.109	(0.049)		41.550	36.990	3.620	5.010	(4.210)	
SA	57	0.381	1.396	–0.094	–0.629	0.383		–0.286	–1.317	–0.905	–1.113	–0.274	
Mean Diff.		(0.118)	(0.111)	0.042	0.177	(0.056)		1.640	(7.58)	5.210	6.430	(1.580)	
Panel B: year 2008–2009													
ALL	330	0.100	0.533	–0.267	–0.167	–0.299		–0.235	–1.047	–0.075	–0.127	–0.416	
Mean Diff.		(0.016)	(0.024)	0.059	0.027	(0.765)		(3.680)	(16.490)	(1.190)	2.010	(6.540)	
ME	135	–0.416	0.021	–0.597	–0.888	0.021		–0.373	–0.418	–0.025	–0.848	–0.963	
Mean Diff.		0.113	(0.001)	0.238	0.225	(0.004)		3.620	4.040	(0.240)	8.210	(9.310)	
SEA	137	0.090	–0.392	0.501	0.457	0.073		–0.615	–0.268	–0.386	–0.694	–0.015	
Mean Diff.		(0.025)	0.039	(0.184)	(0.139)	(0.009)		(7.020)	(3.100)	(4.460)	(8.030)	(0.170)	
SA	58	0.362	2.168 ^c	–0.374	0.425	–1.155		–0.588	–2.419 ^b	–0.401	–0.227	–0.093	
Mean Diff.		(0.073)	(1.147)	0.079	(0.088)	0.129)		(3.440)	(13.200)	2.360	(1.330)	0.540	
Panel C: year 2009–2010													
ALL	341	3.075 ^a	0.796	2.538 ^b	1.086	0.174		–1.887 ^c	–0.318	–1.384	–1.032	–1.355	
Mean Diff.		(0.368)	(0.026)	(0.332)	(0.229)	(0.021)		(27.960)	(4.720)	(20.550)	(15.330)	(20.11)	
ME	146	2.601 ^b	0.353	2.423 ^b	1.980 ^b	–0.414		–1.530	–0.463	–1.407	–1.036	–0.870	
Mean Diff.		(0.499)	(0.019)	(0.493)	(0.446)	0.093		(13.990)	4.230	(12.870)	(9.480)	(7.960)	
SEA	139	0.841	0.216	0.742	0.444	0.733		–0.781	–0.350	–0.564	–0.518	–0.093	
Mean Diff.		(0.252)	(0.009)	(0.239)	(0.139)	(0.106)		(8.260)	(3.740)	(6.040)	(5.550)	(0.980)	
SA	56	0.890	0.896	0.533	–0.276	1.125		–0.267	–1.284	–0.214	–0.107	–1.458	
Mean Diff.		(0.323)	(0.052)	(0.202)	0.069	(0.245)		(1.520)	(7.300)	(1.220)	0.610	(8.280)	
Panel D: year 2010–2011													
ALL	344	–0.645	0.141	–0.666	–0.088	0.089		–0.843	–0.426	–0.839	–0.926	–0.299	
Mean Diff.		1.149	(0.004)	0.162	0.019	(0.011)		12.410	(6.290)	12.370	13.640	4.390	
ME	145	–0.161	–0.671	–0.112	0.080	0.450		–0.199	–0.660	–0.330	–0.086	–0.548	
								0.040	0.435	0.109	0.007	0.300	

Table 7 (continued)

Region	No. of Obs.	Parametric test				Nonparametric test								Kruskal–Wallis Equality of Population			
		t test				Mann–Whitney [Wilcoxon Rank Sum] test								χ^2 (Prb > χ^2)			
		t (Prb > t)				z (Prb > z)											
		TPFCH	TECHCH	EFFCH	PECH	SECH	TPFCH	TECHCH	EFFCH	PECH	SECH	TPFCH	TECHCH	EFFCH	PECH	SECH	SECH
Mean Diff.		0.064	0.028	0.045	(0.029)	(0.083)	(1.840)	6.090	(3.050)	0.800	(5.050)	(1.840)	6.090	(3.050)	0.800	(5.050)	
SEA	139	0.214	−0.519	0.486	0.249	0.857	−0.646	−0.289	−0.897	−0.210	−0.913	0.418	0.083	0.805	0.044	0.834	
Mean Diff.		(0.074)	0.029	(0.190)	(0.099)	(0.162)	(6.540)	2.940	(9.120)	(2.140)	(9.280)	(6.540)	2.940	(9.120)	(2.280)	(9.280)	
SA	60	−1.522	2.237 ^b	−2.129 ^b	−1.551	−1.330	−2.761 ^a	−2.168 ^b	−3.210 ^a	−2.637 ^a	−2.596 ^a	7.622 ^a	4.702 ^b	10.302 ^a	6.952 ^a	6.737 ^a	
Mean Diff.		0.623	(0.126)	0.937	0.340	0.365	16.090	(12.640)	18.700	15.370	15.030	16.090	(12.640)	18.700	15.370	15.030	
<i>Panel E: year 2011–2012</i>																	
ALL	349	−0.782	1.740 ^c	−0.986	0.245	−0.910	−0.011	−0.597	−0.371	−0.334	−0.808	0.001	0.356	0.138	0.111	0.652	
Mean Diff.		0.154	(0.043)	0.187	(0.046)	0.256	0.170	(8.840)	5.510	4.940	(11.950)	0.170	(8.840)	5.510	4.940	(11.950)	
ME	145	−0.619	0.269	−0.649	−0.135	−0.680	−0.207	−0.013	−0.450	−0.456	−0.577	0.043	0.001	0.203	0.207	0.333	
Mean Diff.		0.204	(0.011)	0.197	0.042	0.079	1.910	(0.120)	4.160	4.200	(5.320)	1.910	(0.120)	4.160	4.200	(5.320)	
SEA	144	0.224	1.444	0.067	0.341	0.600	−0.145	−0.650	−0.453	−0.643	−0.266	0.021	0.422	0.205	0.414	0.071	
Mean Diff.		(0.057)	(0.080)	(0.016)	(0.112)	(0.107)	1.520	(6.820)	4.770	6.770	(2.790)	1.520	(6.820)	4.770	6.770	(2.790)	
SA	60	−0.733	1.647	−0.919	−0.028	−0.936	−0.083	−0.240	−0.277	−0.111	−1.187	0.007	0.058	0.077	0.012	1.410	
Mean Diff.		0.377	(0.049)	0.503	0.005	1.139	(0.460)	(1.350)	1.560	(0.630)	(6.660)	(0.460)	(1.350)	1.560	(0.630)	(6.660)	
<i>Panel F: year 2012–2013</i>																	
ALL	366	−1.082	−2.465 ^b	−0.756	0.199	−0.984	−0.458	−1.566	−0.067	−0.222	−1.143	0.210	2.452	0.004	0.049	1.307	
Mean Diff.		0.256	0.066	0.161	(0.053)	0.084	6.550	22.380	(0.960)	(3.170)	16.330	6.550	22.380	(0.960)	(3.170)	16.330	
ME	145	−0.938	−1.433	−0.705	−0.130	−0.713	−1.337	−1.155	−0.932	−0.283	−1.138	1.789	1.333	0.868	0.080	1.295	
Mean Diff.		0.317	0.058	0.263	0.074	0.063	(11.820)	10.200	8.230	2.500	10.050	11.820	10.200	8.230	2.500	10.050	
SEA	148	−0.851	−1.193	−0.596	0.467	−1.039	−0.075	−0.040	−0.094	−0.275	−0.771	0.006	0.002	0.009	0.076	0.595	
Mean Diff.		0.429	0.058	0.288	(0.122)	0.189	0.750	0.400	0.940	2.720	7.630	0.750	0.400	0.940	2.720	7.630	
SA	63	0.517	−2.390 ^b	0.940	1.395	−0.139	−1.240	−2.208 ^b	−1.732 ^c	−1.699 ^c	−0.477	1.537	2.885 ^b	3.001 ^c	2.885 ^c	0.228	
Mean Diff.		(0.153)	0.104	(0.282)	(0.298)	0.020	(7.080)	12.600	(9.890)	(9.690)	2.710	(7.080)	12.600	(9.890)	(9.690)	2.710	
<i>Panel G: Year 2013–2014</i>																	
ALL	370	0.003	−0.953	−0.008	−0.051	−0.430	−1.117	−0.276	−1.461	−1.087	−1.318	1.249	0.076	2.133	1.182	1.736	
Mean Diff.		(0.001)	0.025	0.002	0.010	0.057	(16.430)	4.060	(21.470)	(15.980)	(19.350)	(16.430)	4.060	(21.470)	(15.980)	(19.350)	
ME	149	1.573	−1.852 ^c	2.423 ^b	1.000	1.718 ^c	−0.309	−1.458	−1.268	−0.688	−1.691 ^c	0.096	2.125	1.608	0.473	2.860	
Mean Diff.		(0.369)	0.101	(0.600)	(0.379)	(0.184)	(2.760)	13.010	(11.320)	(6.150)	(15.070)	(2.760)	13.010	(11.320)	(6.150)	(15.070)	
SEA	151	−0.830	1.223	−1.055	−0.597	−0.801	−1.217	−0.611	−0.862	−1.157	−0.005	1.481	0.373	0.743	1.340	0.001	
Mean Diff.		0.375	(0.053)	0.627	0.265	0.481	(12.060)	(6.050)	(8.540)	(11.470)	(0.050)	(12.060)	(6.050)	(8.540)	(11.470)	(0.050)	

Table 7 (continued)

Region	No. of Obs.	Parametric test						Nonparametric test						Kruskal–Wallis Equality of Population					
		<i>t</i> test						Mann–Whitney [Wilcoxon Rank Sum] test						χ^2 (Prb > χ^2)					
		<i>t</i> (Prb > <i>t</i>)						<i>z</i> (Prb > <i>z</i>)											
		TFPCH	TECHCH	EFFCH	PECH	SECH		TFPCH	TECHCH	EFFCH	PECH	SECH		TFPCH	TECHCH	EFFCH	PECH	SECH	
SA	70	−0.192	−0.258	−0.258	−1.000	2.691 ^a		−0.717	−0.725	−0.662	−0.304	−1.046		0.514	0.525	0.439	0.092	1.093	
Mean Diff.		0.071	0.013	0.099	0.344	(0.206)		(4.620)	(4.670)	(4.270)	(1.960)	(6.740)		(4.620)	(4.670)	(4.270)	(1.960)	(6.740)	
<i>Panel H: year 2014–2015</i>																			
ALL	369	0.734	3.334 ^a	−0.198	0.659	−1.891 ^c		−0.898	−2.960 ^a	−0.131	−1.139	−2.253 ^b		0.807	8.763 ^a	0.017	1.297	5.077 ^b	
Mean Diff.		(0.146)	(0.104)	0.038	(0.152)	0.197		(13.180)	(43.410)	1.920	(16.700)	33.000		(13.180)	(43.410)	1.920	(16.700)	33.000	
ME	151	1.379	2.695 ^a	0.868	0.941	−0.818		−1.493	−2.846 ^a	−0.727	−0.777	−0.244		2.228	8.100 ^a	0.529	0.604	0.059	
Mean Diff.		(0.481)	(0.135)	(0.288)	(0.395)	0.079		(13.490)	(25.720)	(6.570)	(7.020)	2.200		(13.490)	(25.720)	(6.570)	(7.020)	2.200	
SEA	147	−0.821	0.481	−1.206	−0.864	−0.103		−0.521	−0.003	−0.744	−0.076	−0.704		0.272	0.001	0.553	0.006	0.495	
Mean Diff.		0.239	(0.024)	0.329	0.293	0.013		5.130	0.020	7.330	(0.750)	6.930		5.130	0.020	7.330	(0.750)	6.930	
SA	71	0.262	3.080	−0.583	1.212	−2.104 ^c		−0.572	−2.498 ^b	−0.349	−1.644 ^c	−3.545 ^a		0.328	6.239 ^b	0.122	2.703 ^c	12.568 ^a	
Mean Diff.		(0.098)	(0.191)	0.246	(0.431)	0.796		(3.620)	(15.820)	2.220	(10.410)	22.410		(3.620)	(15.820)	2.220	(10.410)	22.410	
<i>Panel I: year 2015–2016</i>																			
ALL	361	−2.698 ^a	−1.584	−2.478 ^b	−2.076 ^b	−0.349		−3.665 ^a	−1.330	−2.872 ^a	−3.378 ^a	−0.195		13.432 ^a	1.769	8.248 ^a	11.413 ^a	0.038	
Mean Diff.		0.639	0.046	0.645	0.536	0.030		54.440	19.760	42.660	50.170	2.890		54.440	19.760	42.660	50.170	2.890	
ME	148	−2.421 ^b	−1.661 ^c	−2.192 ^b	−1.987 ^b	−0.565		−3.155 ^a	−1.351	−2.515 ^b	−2.887 ^a	−0.027		9.956 ^a	1.824	6.325 ^b	8.337 ^a	0.001	
Mean Diff.		1.076	0.081	1.081	0.914	0.076		28.790	12.320	22.50	26.350	(0.250)		28.790	12.320	22.50	26.350	(0.250)	
SEA	146	−1.449	−0.064	−1.495	−1.844 ^c	0.733		−2.180 ^b	−0.722	−2.172 ^b	−2.429 ^b	−0.107		4.752 ^b	0.521	4.717 ^b	5.900 ^b	0.011	
Mean Diff.		0.356	0.003	0.366	0.439	(0.114)		21.320	7.060	21.250	23.760	(1.050)		21.320	7.060	21.250	23.760	(1.050)	
SA	67	−0.110	−0.631	−0.070	0.873	−0.909		−0.370	−0.070	−0.220	−0.669	−0.397		0.137	0.005	0.048	0.448	0.157	
Mean Diff.		0.044	0.043	0.030	(0.327)	0.202		2.470	0.470	(1.470)	(4.470)	2.640		2.470	0.470	(1.470)	(4.470)	2.640	
<i>Panel J: year 2016–2017</i>																			
ALL	364	0.393	−3.022 ^a	3.753 ^a	2.865 ^b	1.534		−0.225	−2.446 ^b	−1.113	−1.240	−0.184		0.050	5.984 ^b	1.239	1.538	0.034	
Mean Diff.		(0.088)	0.144	(0.980)	0.144	(0.138)		(3.290)	35.880	(16.330)	(18.200)	(2.700)		(3.290)	35.880	(16.330)	(18.200)	(2.700)	
ME	148	2.095 ^b	−2.355 ^b	3.597 ^a	2.226 ^b	0.547		−0.698	−2.142 ^b	−1.469	−1.427	−0.460		0.487	4.588 ^b	2.157	2.037	0.212	
Mean Diff.		(0.512)	0.142	(1.469)	(1.273)	(0.101)		(6.280)	19.280	(13.210)	(12.840)	(1.140)		(6.280)	19.280	(13.210)	(12.840)	(1.140)	
SEA	147	−1.050	−3.341 ^a	1.549	0.721	1.867 ^c		−0.709	−2.552 ^b	−0.176	−0.291	−0.326		0.502	6.513 ^b	0.031	0.084	0.106	
Mean Diff.		0.435	0.275	(0.760)	(0.579)	(0.257)		6.970	25.120	1.740	2.860	(3.210)		6.970	25.120	1.740	2.860	(3.210)	
SA	69	0.510	0.243	0.960	2.222 ^b	−0.117		−0.602	−0.024	−1.005	−1.282	−0.705		0.362	0.001	1.011	1.644	0.496	
Mean Diff.		(0.097)	(0.026)	(0.410)	(0.599)	0.016		(3.840)	0.160	(6.400)	(8.170)	4.490		(3.840)	0.160	(6.400)	(8.170)	4.490	

Table 7 (continued)

Region	No. of Obs.	Parametric test					Nonparametric test					Kruskal–Wallis Equality of Population				
		t test					Mann–Whitney [Wilcoxon Rank Sum] test					$\chi^2(\text{Prb} > \chi^2)$				
		$t(\text{Prb} > t)$					$z(\text{Prb} > z)$									
		TFPCH	TECHCH	EFFCH	PECH	SECH	TFPCH	TECHCH	EFFCH	PECH	SECH	TFPCH	TECHCH	EFFCH	PECH	SECH
Panel K: all years and regions																
ALL	3515	−1.488	−1.700 ^c	0.152	0.765	−1.034	−0.605	−0.408	−0.090	−0.202	−0.044	0.366	0.166	0.008	0.041	0.002
Mean Diff.		0.100	0.019	(0.012)	(0.066)	0.049	28.360	19.100	(4.240)	(9.480)	2.070	28.360	19.100	(4.240)	(9.480)	2.070
ME	1454	0.007	−1.340	0.910	0.737	−0.044	−0.161	−0.847	−0.503	−0.004	−0.779	0.026	0.717	0.253	0.001	0.607
Mean Diff.		(0.000)	0.024	(0.120)	(0.119)	0.002	4.640	24.440	(14.510)	0.110	(22.470)	4.640	24.440	(14.510)	0.110	(22.470)
SEA	1430	−2.014 ^b	−2.957 ^a	−0.328	0.044	−0.060	−0.897	1.945 ^c	−0.105	−0.112	−0.223	0.805	3.783 ^c	0.011	0.012	0.050
Mean Diff.		0.255	0.059	0.044	(0.006)	0.004	29.030	63.000	(3.420)	3.610	(7.240)	29.030	63.000	(3.420)	3.610	(7.240)
SA	631	−0.251	2.160 ^b	−0.790	1.210	−1.629	−0.103	−3.019 ^a	−0.590	−0.776	−1.401	0.011	9.116 ^a	0.348	0.602	0.011
Mean Diff.		0.028	(0.050)	0.099	(0.101)	0.236	(1.950)	(57.140)	11.160	(14.680)	26.470	(1.950)	(57.140)	11.160	(14.680)	26.470

^a, ^b, ^c indicate significance at the 1%, 5% and 10% levels, respectively

In bracket indicates the productivity mean of conventional banks is higher than Islamic banks

Abbreviation

DEA: Data envelopment analysis; TFP: Total factor productivity; TFPCH: TFP change; GFC: Global financial crisis; SA: South Asia; SEA: Southeast Asia; ME: Middle East; SAC: Shariah Advisory Committee; GCC: Gulf Cooperation Council; MPI: Malmquist productivity index; OIC: Organisation of Islamic Cooperation; SFA: Stochastic frontier analysis; TECHCH: Technical change; EFFCH: Efficiency change; SECH: Scale efficiency change; PTECH: Pure technical change; CRS: Constant returns to scale; VRS: Variable returns to scale; DMU: Decision-making unit.

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Authors' contributions

RWWJ structured the theoretical, research framework and analysed the data. FK conceived the idea and clarified the main issues in the research to ensure the novelty contribution and represent as the corresponding author. ARAL collects and provides data for this study. HHH performed the interpretation from the analysis results obtained. KMT ensures all the previous literature cited related to the issues discussed in the study and keeps research flow accordingly. All authors read and approved the final manuscript

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Availability of data and material

The datasets used and/or analysed during the current study are available from the corresponding author on reasonable request.

Declarations

Competing interests

The authors declare that they have no competing interests.

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