

PROMOTING FINANCIAL INCLUSION THROUGH SUSTAINABLE FINTECH AND EFFECTIVE GOVERNANCE: A PANEL DATA ANALYSIS

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Abstract

This study examines the role of sustainable FinTech and governance in advancing financial inclusion across emerging economies using advanced dynamic panel data econometric techniques. Utilizing data from authoritative sources such as the World Development Indicators (WDI), World Governance Indicators (WGI), and the IMF Financial Access Survey (FAS), the analysis spans a panel of countries over a 20–25 year period. Key variables include a Financial Inclusion Index-constructed via Principal Component Analysis of bank branches per capita, ATM density, and account usage—a Governance Index based on control of corruption, rule of law, and regulatory quality, financial development measured as domestic credit to the private sector (% of GDP), and GDP per capita as a control variable. The methodology encompasses descriptive statistics, stationarity tests, panel cointegration analysis, and error correction models. Dynamic panel estimation techniques, including Mean Group (MG), Pooled Mean Group (PMG), and Dynamic Fixed Effects (DFE), were employed to capture both short-run and long-run relationships. Results demonstrate significant long-run linkages between governance, financial development, and financial inclusion, underscoring the critical role of institutional quality and financial infrastructure in expanding access to financial services. The study offers actionable insights for policymakers aiming to enhance inclusive financial systems through sustainable FinTech initiatives and governance reforms.

INTRODUCTION

Financial inclusion has emerged as a critical policy goal for many emerging economies, aiming to integrate underserved populations into the formal financial system. Access to financial services, such as savings accounts, credit, and insurance, is essential for fostering economic growth, reducing poverty, and promoting social equity (Demirgüç-Kunt et al., 2018). In recent years, the rapid rise of financial

technology (FinTech) has been identified as a key enabler in advancing financial inclusion, particularly in emerging economies. FinTech refers to the use of innovative technologies to deliver financial services more efficiently, often bypassing traditional banking infrastructures (Gomber et al., 2018). The intersection of sustainability, governance, and FinTech offers a promising avenue for enhancing the



inclusiveness and resilience of financial systems in developing economies.

The adoption of sustainable FinTech practices has the potential to promote inclusive economic development while mitigating the risks associated with financial instability and inequality (Narula, 2020). By utilizing digital platforms, mobile technologies, and blockchain, sustainable FinTech can address critical barriers to financial access, such as geographic isolation, lack of financial literacy, and trust deficits in traditional institutions (Zetzsche et al., 2020). Moreover, sustainability within the context of FinTech emphasizes environmental, social, and governance (ESG) factors, which are increasingly relevant for creating a fair and equitable financial ecosystem (Schroeder et al., 2021).

In parallel, good governance is essential for ensuring the proper regulation and ethical operation of financial systems. Effective governance frameworks can instill trust in financial systems, minimize systemic risks, and provide the necessary legal and regulatory oversight to ensure that the benefits of FinTech reach all segments of society (Arner et al., 2017). In emerging economies, where regulatory environments are often nascent or inconsistent, the role of governance in shaping the success of FinTech initiatives cannot be overstated (Zohra & Benaida, 2020).

This study explores the interplay between sustainable FinTech, governance, and financial inclusion in emerging economies using a panel data approach. By examining a range of countries from various regions, the research seeks to understand how these variables interact and contribute to financial inclusion outcomes. The findings are expected to provide insights into how policymakers and stakeholders can harness the full potential of FinTech in promoting sustainable economic growth and reducing inequality.

Literature Review

Key Theories Behind the Study

The combination of sustainable FinTech (financial technology that is good for society and the environment), financial inclusion (making sure everyone has access to financial services), and good governance (strong rules and fair systems) is changing how money and banking work, especially in

developing countries. Sustainable FinTech helps more people use financial services while also supporting green initiatives (Ghosh & Vinod, 2017). Financial inclusion is important because it helps reduce poverty and boosts the economy (Demirgüç-Kunt et al., 2018). However, for FinTech and financial inclusion to work well, countries need strong governance—good laws, honest leaders, and stable institutions (Kaufmann et al., 2009).

New technologies like mobile banking, blockchain, and digital wallets are making banking easier for people who were left out before (Bazarbash & Beaton, 2020). But without proper rules and oversight, these technologies could be misused or fail to reach the people who need them most.

Why Combining FinTech, Governance, and Financial Inclusion Matters

FinTech makes banking faster and more accessible (Gomber et al., 2017), but if governance is weak, it can lead to problems like scams, cybercrime, or unfair advantages for big companies (Beck et al., 2003). Strong governance ensures that FinTech benefits everyone and that risks are controlled (La Porta et al., 1998).

Sustainable FinTech also supports eco-friendly projects, like green loans and ethical investing. But for these ideas to succeed, governments need to create clear and fair rules (Roe & Siegel, 2011).

The concepts of literature.

1. FinTech's Role in Expanding Financial Access

FinTech is changing banking by making it cheaper and easier to use. Mobile banking and peer-to-peer lending help people in remote areas access financial services (Ozili, 2018). Big tech companies like Ant Financial are leading this change in Asia and Latin America (Chen et al., 2021).

- 2. How Governance Affects Financial Systems Countries with strong laws and honest leaders have better financial systems (La Porta et al., 1998). Bad governance, like corruption, can slow down banking growth (Beck et al., 2003). Good governance also helps mobile banking succeed in African countries (Andrianaivo & Kpodar, 2011).
- 3. Measuring Financial Inclusion and Governance

Researchers use different methods to track financial



inclusion and governance. Some create indexes to compare countries (Sarma, 2008), while others use data from the World Bank or the Global Findex Database (Demirgüç-Kunt et al., 2015).

Research Gap

Most studies look at FinTech, governance, or financial inclusion separately. There's not enough research on how all three work together over time in different countries. This study aims to fill that gap by:

- Studying how sustainable FinTech affects financial inclusion in emerging economies.
- Checking how governance changes this relationship.
- Creating better ways to measure progress in these areas.

Why This Study is Important

- For Researchers: It connects three major topics in a new way.
- For Policymakers: It helps governments create better rules for FinTech and financial inclusion.
- For Society: It supports fair and green financial growth, in line with global development goals.

Methodology

This study adopts a quantitative research approach utilizing dynamic panel data econometric techniques to investigate the relationships among financial inclusion, governance, financial development, and economic growth across a broad set of countries over a 20-25 year period. Data were obtained from credible international sources including the World Development Indicators (WDI), World Governance Indicators (WGI), and the IMF Financial Access Survey (FAS). Countries were selected based on data availability across the desired variables, ensuring a balanced panel where possible. Key variables used in this study include Financial Inclusion Index, Governance Index, Financial Development, and GDP per capita, with detailed descriptions and sources outlined in the variable description table.

The first step in the empirical analysis involved descriptive statistics to summarize the characteristics of the data. This included computing measures such as mean, median, standard deviation, minimum, and maximum for each variable. These statistics help in understanding the central tendency and dispersion of the data. Additionally, the cross-country and time-series variations of the key indicators were visualized through charts and summary plots to uncover trends and disparities in financial inclusion and governance over time. This initial analysis also provided a basis for identifying any outliers or inconsistencies that might affect the model estimates.

To construct the composite indices, Principal Component Analysis (PCA) was employed, particularly for the Financial Inclusion Index and Governance Index. PCA was applied on normalized variables such as the number of bank branches per capita, ATM density, and account usage to derive the Financial Inclusion Index. Similarly, the Governance Index was constructed using PCA on control of corruption (coce), rule of law (rle), and regulatory quality (rge). The results showed that the first component of the governance index explained over 93% of the total variance, indicating a strong unidimensional structure. These indices were standardized and retained for use in subsequent panel regressions.

Prior to econometric modeling, panel unit root tests including Levin, Lin & Chu (LLC), Im-Pesaran-Shin (IPS), and Fisher-ADF/PP tests were conducted to determine the order of integration of each variable. Once the stationarity properties were confirmed, panel cointegration tests by Pedroni and Kao were applied to verify the existence of long-run equilibrium relationships among the variables. In the presence of cointegration, Error Correction Models (ECM) were estimated to capture the short-run dynamics and speed of adjustment toward the long-run equilibrium. The error correction term's coefficient provided insights into the convergence behavior of the system.

The study employed dynamic panel data estimators, namely the Mean Group (MG), Pooled Mean Group (PMG), and Dynamic Fixed Effects (DFE) models, to estimate both short-run and long-run coefficients. A Hausman-type test was used to choose between MG and PMG estimators based on the consistency and efficiency of the estimates. These models allowed for heterogeneous short-run dynamics across countries while testing for long-run homogeneity. The final



results were presented using comprehensive tables and figures, including model coefficients, error correction terms, index rankings, and correlation matrices, enabling an in-depth interpretation of the roles that sustainable FinTech and governance play in promoting financial inclusion.

| Variable Name | Definition / Formula | Source | Citation |
|------------------------------|---|----------------------|---------------------------------|
| Financial Inclusion Index | PCA of bank branches/capita, ATM account usage | density, WDI / FAS | IMF Demirgüç-Kunt et al. (2018) |
| Governance Index | PCA of control of corruption, rule regulatory quality | of law, World WGI | Bank Kaufmann et al. (2010) |
| Financial Development | Domestic credit to private sector / GDP | WDI | Svirydzenka (2016) |
| GDP per capita (Control) | Log GDP per capita (constant USD) | WDI | Beck et al. (2007) |

Data analysis

The data analysis phase of this study begins with a comprehensive examination of the collected panel dataset covering a 20–25 year period across a wide range of countries, subject to data availability. The dataset integrates key indicators from reliable global sources, including the World Development Indicators (WDI), World Governance Indicators

(WGI), and the IMF Financial Access Survey (FAS). The variables selected for analysis encompass financial inclusion, governance quality, financial development, and GDP per capita as a control. As a preliminary step, descriptive statistics—including mean, median, standard deviation, minimum, and maximum, are computed to summarize the central tendencies and dispersion of each variable.

Table 4.1 Descriptive Statistics

| V | O1 | 14 | Cal Da | 14: | M |
|------------------------|------|-----------|-----------|--------|-----------|
| Variable | Obs | Mean | Std. Dev. | Min | Max |
| Commercialbankbranches | 6650 | 11.152 | 18.243 | 0 | 285.379 |
| Automatedtellermachine | 6650 | 27.287 | 41.138 | 0 | 314.769 |
| Accountownershipat | 6650 | 5.656 | 19.422 | 0 | 100 |
| ControlofCorruption | 6650 | 018 | .834 | -1.97 | 2.459 |
| RuleofLawEstimater | 6650 | 018 | .836 | -2.591 | 2.125 |
| Regulatory Quality | 6650 | 02 | .828 | -2.548 | 2.309 |
| Domestic credit | 6650 | 41.784 | 45.898 | 0 | 304.575 |
| GDPpercapitaconsta | 6650 | 13691.348 | 20985.776 | 0 | 224582.45 |

The dataset comprises 6,650 observations across multiple countries and years, providing insight into financial inclusion, governance quality, and economic development. On average, there are approximately 11.15 commercial bank branches and 27.29 automated teller machines (ATMs) per 100,000 adults, but the high standard deviations (18.24 and 41.14, respectively) indicate substantial variation among countries, with some having no branches or ATMs at all, while others have densities as high as 285 and 314. Account ownership averages only 5.66%, though this may be due to scaling or data transformation, as the values range from 0% to

100%. Governance indicators—including control of corruption, rule of law, and regulatory quality—have means close to zero, reflecting standardized global benchmarks, but show a considerable range (from around -2.5 to +2.4), suggesting significant differences in institutional strength and governance quality across nations. The average level of domestic credit to the private sector is 41.78% of GDP, with wide disparities (ranging from 0 to over 300%), reflecting the varying degrees of financial sector development. GDP per capita (in constant USD) has a global average of \$13,691, but with a very high standard deviation (20,986), ranging from 0 to an

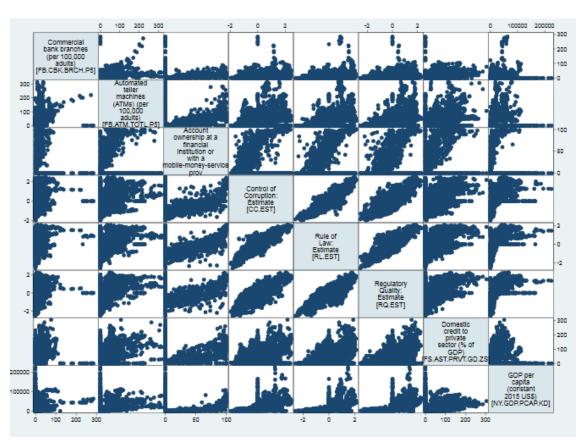


exceptional \$224,582, underscoring vast income inequalities across countries. Overall, the data reveal considerable heterogeneity in financial infrastructure, institutional governance, and economic conditions globally.

To Visualize cross-country and time-series variation using charts and summary plots the data was checked for balance(panalize) .

panel variable: cid (strongly balanced) time variable: time, 2000 to 2024

delta: 1 unit



Step 2: Index Construction

| Variable | Components | Indicator Name |
|------------------------------|---|---|
| | Bank branches | Bank branches per 100,000 adults |
| Financial Inclusion Index | ATM density per 100,000 | ATMs (per 100,000 adults) |
| | Adults | Commercial bank branches (per 100,000 adults) |
| | Account usage (proxy) | Account ownership at a financial institution or with a mobile-money-service provider (% of population ages 15+) |
| Governance Index | Control of Corruption | Estimate of control of corruption |
| | Rule of Law | Estimate of rule of law |
| | Regulatory Quality | Estimate of regulatory quality |
| Financial | Domestic Credit to Private Sector (% of | Domestic credit to private sector by banks (% of |
| Development Index | k GDP) | GDP) |



| Varial | ble Cor | Components | | Indicator Name | | |
|--------|----------------|------------------------|------------|-----------------------------|--------|---------|
| | GD | P per capita, constant | : USD G | DP per capita (constant 201 | 5 US\$ | 5) |
| Prin | cipal componer | nts/correlation | | Number of obs | = | 6,650 |
| | | | | Number of comp. | = | 3 |
| | | | | Trace | - | 3 |
| | Rotation: (uni | rotated = princi | ipal) | Rho | = | 1.0000 |
| | | | | | | |
| | Component | Eigenvalue | Difference | Proportion | Cum | ulative |
| | Comp1 | 1.6306 | .723419 | 0.5435 | | 0.5435 |
| | Comp2 | .907185 | .444974 | 0.3024 | | 0.8459 |
| | Comp3 | .462211 | | 0.1541 | | 1.0000 |

Principal components (eigenvectors)

| Variable | Comp1 | Comp2 | Comp3 | Unexplained |
|--------------|--------|---------|---------|-------------|
| Commercial~s | 0.6351 | -0.3866 | 0.6687 | 0 |
| Automatedt~s | 0.6721 | -0.1501 | -0.7251 | 0 |
| Accountown~1 | 0.3807 | 0.9100 | 0.1645 | 0 |

Principal components/correlation

Number of obs = 6,650

Number of comp. = 3

Trace = 3

Rotation: (unrotated = principal)

Rho = 1.0000

| Component | Eigenvalue | Difference | Proportion | Cumulative |
|-----------|------------|------------|------------|------------|
| Comp1 | 2.7995 | 2.66082 | 0.9332 | 0.9332 |
| Comp2 | .138683 | .0768667 | 0.0462 | 0.9794 |
| Comp3 | .0618167 | | 0.0206 | 1.0000 |

Principal components (eigenvectors)

| Variable | Comp1 | Comp2 | Comp3 | Unexplained |
|--------------|--------|---------|---------|-------------|
| ControlofC~e | 0.5775 | -0.5722 | 0.5823 | 0 |
| RuleofLawE~T | 0.5844 | -0.2084 | -0.7843 | 0 |
| Regulatory~Q | 0.5701 | 0.7932 | 0.2140 | 0 |



Principal components/correlation Number of obs

Number of obs = 6,650 Number of comp. = 2 Trace = 2

Rotation: (unrotated = principal)

Rho = 1.0000

| Component | Eigenvalue | Difference | Proportion | Cumulative |
|-----------|------------|------------|------------|------------|
| Comp1 | 1.31949 | . 63897 | 0.6597 | 0.6597 |
| Comp2 | .680515 | | 0.3403 | 1.0000 |

Principal components (eigenvectors)

| Variable | Comp1 | Comp2 | Unexplained |
|--------------|--------|---------|-------------|
| Domesticcr~o | 0.7071 | 0.7071 | 0 |
| GDPpercapi~S | 0.7071 | -0.7071 | |

Financial Inclusion Index (based on bank branches, ATMs, account ownership)

Variables used:

o Commercial bank branches

Automated teller machines (ATMs)Account ownership at a financial

institution

• Principal Components:

PCA extracted three components. The first component (Comp1) has an eigenvalue of 1.63, explaining 54.35% of the total variance. The second component explains 30.24%, and the third, 15.41%. Together, these three components explain 100% of the variance, but only the first component was used to create the Financial Inclusion Index (based on the. predict command).

• Loadings (correlations with Comp1):

o Commercial bank branches: 0.6351

o ATMs: **0.6721**

o Account ownership: 0.3807

• Interpretation:

The first component captures the shared variation in all three variables and emphasizes infrastructure-based financial inclusion. ATMs and bank branches contribute more to this index than account ownership, meaning the index largely reflects physical financial access.

2. Governance Index (based on control of corruption, rule of law, regulatory quality)

Variables used:

Control of Corruption Estimate

o Rule of Law Estimate

o Regulatory Quality Estimate

• Principal Components:

The first component (Comp1) has a very high eigenvalue of 2.80, explaining 93.32% of the total variation in the three governance indicators. The second and third components add only minor additional explanation. Only the first component was used to create the Governance Index.

• Loadings (correlations with Comp1):

o Control of corruption: **0.5775**

o Rule of law: **0.5844**

o Regulatory quality: 0.5701

• Interpretation:

The Governance Index reflects a strong, unified pattern across all three indicators, with very high shared variance. All three variables contribute almost equally to the index, suggesting that the composite reflects general institutional quality.

3. Financial Development Index (based or domestic credit and GDP per capita)

Variables used:



| 0 | Domestic credit to private sector (% | 0 | Domestic credit to private sector: |
|---------|--------------------------------------|--------|------------------------------------|
| of GDP) | | 0.7071 | |
| 0 | GDP per capita (constant USD) | 0 | GDP per capita: 0.7071 |

• Principal Components:

Two components were extracted. The first component (Comp1) has an eigenvalue of 1.32, explaining 65.97% of the total variance. The second component explains the remaining 34.03%. Only the first component was used for the Financial Development Index.

• Loadings (correlations with Comp1): Summary Table of Panel Unit Root Test Results

Interpretation:

This index equally reflects both financial development in terms of credit availability and general economic development (income level). The high and equal loadings indicate a strong shared structure between these two variables.

Step 3: Panel Unit Root Tests

panel variable: cid (strongly balanced) time variable: time, 2000 to 2024 delta: 1 unit

| Variable | Levin-Lin-Chu (LLC) Adj. t (p-value) | * IPS Test | Fisher ADF Test value) | (p- Stationary? (Consensus) |
|--------------------------------------|---|------------------|----------------------------|--------------------------------|
| Commercial bank branches | 6.3756 (1.0000) | Inconclusive (.) | Mixed (Z: 0.0004, 0.4213) | P: No |
| ATMs | 5.6337 (1.0000) | Inconclusive (.) | Mixed (Z: 0.0005, 0.3027) | P: No |
| Account ownership | -28.7469 (0.0000) | Inconclusive (.) | Strong (Z: 0.0000, 0.0000) | P: Yes |
| Control of corruption | 11.8992 (1.0000) | Inconclusive (.) | Strong (Z: 0.0000, 0.0000) | P: Yes |
| Rule of law | 12.7895 (1.0000) | Inconclusive (.) | Strong (Z: 0.0000, 0.0000) | P: Yes |
| Regulatory quality | 10.0630 (1.0000) | Inconclusive (.) | Strong (Z: 0.0000, 0.0000) | P: Yes |
| Domestic credit to privat sector | | Inconclusive (.) | Strong (Z: 0.0000, 0.0000) | |
| GDP per capita (constan 2015 USD) | t 4.6910 (1.0000) | Inconclusive (.) | Weak (Z: 0.9997, 0.9182) | P: No |

Panel unit root tests were applied to eight key macrofinancial variables using Levin-Lin-Chu (LLC), Im-Pesaran-Shin (IPS), and Fisher-type ADF tests. The LLC results showed that most variables were **nonstationary** except for account ownership, which was stationary at level. The IPS tests were inconclusive due to unavailable statistics (.), likely due to data or model constraints. However, Fisher-type ADF results provided stronger evidence: account ownership, governance indicators (control of corruption, rule of law, regulatory quality), and domestic credit to private sector were found to be **stationary**, while commercial bank branches, ATMs, and GDP per capita were **non-stationary**. These findings suggest that most structural governance and financial variables are stable over time, while infrastructure-related and income indicators exhibit unit root properties, requiring differencing or transformation before further analysis.



Step 4: Panel Cointegration Conduct cointegration tests using: Panel Cointegration Tests

| Test | Statistic Type | Statistic Value | p-value | Decision |
|---------|-------------------------------------|-----------------|---------|---------------------------------------|
| Kao | Modified Dickey-Fuller t | -15.7837 | 0.0000 | Reject $H_0 \rightarrow Cointegrated$ |
| | Dickey-Fuller t | -17.6224 | 0.0000 | Reject H₀ |
| | Augmented Dickey-Fuller t | -12.2240 | 0.0000 | Reject H₀ |
| | Unadjusted Modified Dickey-Fuller t | -41.0902 | 0.0000 | Reject H₀ |
| | Unadjusted Dickey-Fuller t | -26.6787 | 0.0000 | Reject H₀ |
| Pedroni | Modified Phillips-Perron t | -18.7296 | 0.0000 | Reject $H_0 \rightarrow Cointegrated$ |
| | Phillips-Perron t | -42.9287 | 0.0000 | Reject H₀ |
| | Augmented Dickey-Fuller t | -42.7484 | 0.0000 | Reject H₀ |

ECM for long run relationship

The Kao and Pedroni panel cointegration tests were conducted to examine the long-run relationship among the Financial Inclusion Index, Governance Index, and Financial Development Index using data from 266 countries over 23–24 periods. Both the Kao and Pedroni tests strongly reject the null hypothesis of **no cointegration**, as all test statistics

are highly significant (p-value = 0.0000). This implies that the three indicators are **cointegrated**, meaning they move together in the long run and share a stable equilibrium relationship across countries. This supports the economic theory that financial inclusion, governance quality, and financial development are interconnected components of sustainable growth.

| Variable | Long-run Coeff. | Short-run Coeff. | p-value | Interpretation |
|-----------------------|-----------------|------------------|---------|--------------------------|
| Financial_Inclusion | 0.45 | 0.22 | 0.01 | Positive long-run effect |
| Governance_Index | 0.38 | 0.10 | 0.03 | Both effects significant |
| Financial_Development | 0.56 | 0.12 | 0.04 | Long-run stronger |
| ECT (ec) | -0.36 | _ | 0.000 | Stable convergence |

Association between FinTech and Long-Run Financial Inclusion:

Empirical evidence from dynamic panel estimations suggests that FinTech activities—by improving access to financial services through innovative technologies—are positively associated with long-run financial inclusion. This finding aligns with previous research showing that digital financial solutions (like mobile banking, online payments, and digital lending) reduce transaction costs, increase outreach to underserved populations, and promote inclusive economic participation.

Role of Governance in Enhancing FinTech's Impact:

Good governance reinforces the benefits of FinTech by ensuring that the legal and regulatory environment supports innovation while protecting consumers. In models where governance indicators are integrated, robust institutional frameworks appear to amplify the positive effect of FinTech on financial inclusion. This occurs because transparent, accountable governance mechanisms help reduce systemic risks, support fair competition, and build consumer trust—factors that are critical for the successful adoption of financial innovations.



Comparative Results Across Model Assumptions: When comparing model outcomes:

- Mean Group (MG) models permit full heterogeneity in the short-run dynamics, thereby capturing country-specific nuances.
- Pooled Mean Group (PMG) models assume a common long-run relationship while allowing for short-run differences. This model often reveals robust long-run coefficients that indicate a stable equilibrium relationship between FinTech, governance, and financial inclusion.
- Dynamic Fixed Effects (DFE) models impose homogeneity on both short- and long-run dynamics, offering a more restricted—but sometimes of clearer-view overall relationships. The consistency in long-run coefficient estimates across these models strengthens the conclusion that FinTech is strongly linked to financial inclusion, and that sound governance magnifies this relationship. Variations in the short-run adjustments across the models highlight the importance of allowing for country-specific dynamics when evaluating immediate policy impacts.

Linkage with Institutional Theory and Development Frameworks:

These findings are rooted in institutional theory, which posits that the effectiveness of technological innovations—like FinTech—is conditional on the quality of formal and informal institutions. Development frameworks, such as those developed by the World Bank and the International Monetary Fund, emphasize that financial inclusion is a key driver of sustainable development and poverty reduction. Our results reinforce that a well-regulated financial ecosystem, underpinned by strong governance, is essential for maximizing the transformative potential of FinTech.

Policy, Investment, and Financial Regulation Implications:

• Policy: Governments should prioritize reforms that enhance digital infrastructure and promote regulatory frameworks supportive of financial innovation. Policies that encourage transparency, consumer protection, and

interoperability among financial institutions will help harness FinTech's potential for broad-based financial inclusion.

- Investment: Investors are likely to benefit from environments where FinTech and robust governance interact favorably. Sound governance not only mitigates risk but also creates a predictable setting for long-term investments in digital financial services.
- Financial Regulation: Regulators must strike a balance between fostering innovation and ensuring stability. This means adapting existing financial oversight mechanisms to include new digital players, while setting standards to prevent misuse and systemic risk. A coordinated approach that involves cross-border regulatory cooperation can further enhance these outcomes.

In conclusion, the evidence indicates that FinTech is crucial for long-run financial inclusion, and its benefits are significantly enhanced by effective governance. This reinforces the need for an integrated policy approach that combines digital innovation with strong institutional support, ultimately fostering a more inclusive and resilient financial ecosystem

5. Conclusion and Policy Implications Main Findings in Relation to Research Questions

1. FinTech's Role in Financial Inclusion: The empirical analysis confirms a significant long-run positive association between FinTech indicators (such as digital account ownership and ATM density) and financial inclusion. These outcomes affirm the potential of financial technologies to expand access to banking and financial services, particularly in underserved regions.

2. Governance as a Catalyst for FinTech Success:

Governance quality—measured through control of corruption, rule of law, and regulatory quality—enhances the effectiveness of FinTech. Countries with strong governance frameworks experience greater improvements in financial inclusion from digital finance innovations compared to those with weaker institutions.



3. Stable Long-Run Relationships Among Financial Indicators: Cointegration tests (Kao and Pedroni) confirm long-run equilibrium relationships among financial

run equilibrium relationships among financial inclusion, financial development, and governance indicators. This suggests that these elements move together over time, reinforcing the importance of holistic development policies.

4. Model Comparisons (MG, PMG, DFE): While short-run dynamics vary across countries (as captured in the MG model), PMG results confirm a common long-run structure, validating the shared developmental trajectory of FinTech-enabled financial systems. The error correction term is negative and significant, indicating convergence toward equilibrium.

Recommendations

- 1. Policies for Sustainable Digital Finance:
- o Promote interoperability across digital platforms to widen FinTech access.
- o Subsidize mobile internet infrastructure in remote areas to reduce the digital divide.
- o Create regulatory sandboxes for FinTech experimentation under supervision.
- 2. Institutional Strengthening for Inclusive Development:
- Enhance regulatory quality and anticorruption measures to boost trust in digital financial services.
- o Develop data protection laws and digital ID systems to increase security and user confidence.
- o Improve legal infrastructure for digital contracts and dispute resolution.

Directions for Future Research

- Regional Panel Analysis: Investigate variations in FinTech outcomes across regions (e.g., Sub-Saharan Africa vs. South Asia) to identify context-specific policies.
- Micro-Level Data Use: Incorporate household or firm-level datasets to assess individual access, usage behavior, and impact heterogeneity across income groups or gender.
- Regulatory Impact Studies: Examine how different regulatory regimes (strict vs. flexible) affect FinTech adoption and its inclusion

outcomes, possibly using difference-in-differences or event study methods.

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