

STATISTICAL APPROACHES IN FORECASTING FOREIGN DIRECT INVESTMENT FLOWS

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Abstract: Foreign Direct Investment (FDI) plays a critical role in the economic development and globalization of national economies. Accurate forecasting of FDI flows is essential for policymakers, investors, and economic planners. This paper explores the application of advanced statistical methods in forecasting FDI inflows and outflows. The study reviews traditional econometric models such as Autoregressive Integrated Moving Average (ARIMA), Vector Autoregression (VAR), and Cointegration Analysis, as well as modern approaches including Machine Learning algorithms and Bayesian forecasting models. Empirical data are analyzed to compare the forecasting accuracy of these models across various economic conditions and countries. The results highlight the strengths and limitations of each method, demonstrating that hybrid models combining statistical and machine learning techniques often yield higher forecasting precision. The paper concludes with policy implications and recommendations for enhancing FDI forecasting practices through the integration of statistical innovations and real-time data analytics.

Keywords: Foreign Direct Investment (FDI), Forecasting, Econometrics, ARIMA, VAR, Cointegration, Machine Learning, Bayesian Models, Statistical Modeling, Economic Forecasting, Hybrid Models, Data Analytics.

Introduction

Foreign Direct Investment (FDI) represents a key driver of economic development, particularly for emerging and transition economies such as Uzbekistan. FDI not only brings much-needed capital but also facilitates the transfer of technology, managerial expertise, employment opportunities, and integration into global value chains. In the post-independence period, Uzbekistan has gradually shifted from a state-controlled economy toward market liberalization, and FDI has played an increasingly important role in supporting this transition.

In recent years, Uzbekistan has implemented significant reforms aimed at improving the business climate, simplifying tax regulations, liberalizing the currency market, and reducing administrative barriers for foreign investors. The Government's "Open Uzbekistan" policy, membership in various international organizations, and the negotiation of investment agreements have further elevated the country's attractiveness for foreign investors. According to the State Committee of the Republic of Uzbekistan on Statistics, FDI inflows have demonstrated positive trends, especially after 2017, when comprehensive economic liberalization policies were initiated.

Despite these positive developments, forecasting FDI flows remains a complex and challenging task due to the multidimensional nature of FDI determinants. Various

macroeconomic, institutional, political, and global factors influence investment decisions, and their interplay is not always linear or easily predictable. Therefore, statistical forecasting methods offer a valuable tool to better understand, model, and predict FDI trends, allowing policymakers and investors to make more informed decisions.

The ability to forecast FDI flows is crucial for several reasons:

- **Policy Planning and Economic Stability:** Reliable forecasts allow governments to plan infrastructure development, resource allocation, and sectoral investment priorities more effectively.
- **Private Sector Decision-Making:** Domestic and foreign firms can assess investment risks and opportunities based on expected FDI trends.
- **International Competitiveness:** Accurate forecasts can help Uzbekistan benchmark its performance against regional competitors and adjust its investment promotion strategies accordingly.
- **Macroeconomic Forecasting:** Since FDI influences exchange rates, employment, production, and fiscal revenues, its accurate prediction contributes to more comprehensive macroeconomic forecasting models.

Given these important implications, selecting appropriate statistical approaches for FDI forecasting becomes an essential academic and policy task.

Forecasting FDI flows in Uzbekistan presents several unique challenges:

- **Limited Historical Data:** Compared to developed economies, Uzbekistan's available time series data on FDI is relatively short and sometimes inconsistent due to previous reporting standards.
- **Structural Economic Changes:** Frequent policy shifts, regulatory reforms, and political changes can suddenly alter the investment environment.
- **External Shocks:** Global financial crises, geopolitical tensions, and commodity price fluctuations directly impact FDI flows into emerging markets like Uzbekistan.
- **Sectoral Shifts:** Uzbekistan is diversifying its economy from traditional sectors such as cotton and gas exports to more technology-intensive, service-oriented, and industrial sectors, requiring updated models that can capture sectoral FDI dynamics.

These factors underscore the need to adopt flexible and sophisticated statistical approaches capable of accommodating structural breaks, non-linearity, and dynamic interdependencies.

Statistical methods for forecasting FDI flows can be broadly categorized into traditional econometric models and modern data-driven approaches.

- **ARIMA (Autoregressive Integrated Moving Average):** ARIMA models are widely used for time series forecasting where data exhibit autocorrelation but no external explanatory variables. ARIMA can effectively capture short-term dynamics but may fail to incorporate structural factors influencing FDI.
- **VAR (Vector Autoregression):** VAR models allow for multiple interrelated time series to be modeled together, capturing feedback relationships between FDI and macroeconomic indicators such as GDP growth, exchange rates, inflation, and interest rates.

- Cointegration and Error Correction Models (ECM): Cointegration techniques assess long-run equilibrium relationships between FDI and its determinants, while ECM corrects short-term deviations, offering a comprehensive dynamic modeling approach.

- Granger Causality Tests: These tests examine the predictive causality relationships between FDI and other macroeconomic variables.

- Artificial Neural Networks (ANN): Neural networks can model complex, non-linear relationships and are increasingly applied in FDI forecasting, especially when large datasets are available.

- Support Vector Machines (SVM): SVMs are effective for regression tasks involving non-linear patterns in FDI determinants.

- Random Forest and Ensemble Methods: These models combine multiple decision trees to improve prediction accuracy and reduce overfitting.

- Bayesian Forecasting: Bayesian models offer probabilistic forecasting that accounts for model uncertainty, which is particularly useful when data is limited or subject to frequent shocks.

- Hybrid Models: Combining traditional econometrics with machine learning allows for capturing both linear and non-linear aspects of FDI dynamics.

Each of these approaches has strengths and weaknesses depending on data availability, forecasting horizon, model complexity, and policy relevance.

Several empirical studies and policy reports have identified key determinants of FDI inflows to Uzbekistan, including:

- Macroeconomic Stability: Inflation rate, GDP growth, exchange rate stability.

- Regulatory Environment: Ease of Doing Business Index, tax reforms, legal protections for investors.

- Political Stability: Absence of conflict, regulatory transparency, and political predictability.

- Infrastructure Development: Availability of transport, energy, and communication facilities.

- Human Capital: Education level, labor market flexibility, and skill availability.

- Global Factors: Commodity prices, global interest rates, and regional economic integration (e.g., participation in EAEU, WTO accession process, and cooperation with China's Belt and Road Initiative).

In recent years, Uzbekistan has made substantial improvements in many of these areas, but vulnerabilities remain in others, such as banking sector reforms, judicial independence, and bureaucratic efficiency.

While several studies have explored FDI determinants globally and in the Central Asian region, there is limited empirical research that applies advanced statistical forecasting techniques specifically to the Uzbek context. Most existing studies rely on descriptive analysis or simple regression models, which may not fully capture the dynamic and non-linear nature of FDI flows.

This research aims to fill this gap by:

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- Applying both traditional econometric and modern machine learning methods to forecast FDI inflows to Uzbekistan.
- Comparing the predictive performance of different statistical models using empirical data from Uzbekistan.
- Offering policy-relevant recommendations based on forecast results to strengthen Uzbekistan's investment promotion strategies.

Literature Review

Foreign Direct Investment (FDI) forecasting is a complex area that lies at the intersection of international economics, statistics, econometrics, and policy studies. Numerous theoretical models and empirical studies have been developed to explain and predict FDI flows across countries and regions. This literature review provides a comprehensive examination of key theories, empirical findings, and statistical approaches relevant to forecasting FDI, with particular emphasis on their application to emerging economies such as Uzbekistan.

Several economic theories attempt to explain why multinational enterprises (MNEs) engage in FDI rather than trade or contractual relationships. These theories form the basis for identifying explanatory variables in empirical forecasting models.

John Dunning's Eclectic Paradigm (1980) remains one of the most widely cited frameworks in FDI research. The model suggests that FDI occurs when firms possess:

- Ownership advantages (O): Firm-specific assets such as technology, brand, or managerial expertise.
- Location advantages (L): Host country factors such as natural resources, labor costs, infrastructure, and political stability.
- Internalization advantages (I): Benefits of controlling foreign operations rather than contracting them to local firms.

In the context of Uzbekistan, location advantages such as mineral resources, favorable geographic position, and labor costs have historically attracted FDI, while internalization incentives have become stronger due to improved legal protections for foreign investors.

Hymer (1976) and Kindleberger (1969) argue that imperfections in capital markets, information asymmetries, and barriers to trade can make direct investment a more efficient way for firms to expand internationally. This theory underscores the importance of reducing transaction costs and improving institutional quality to attract FDI.

Gravity models, widely applied in international trade, have been adapted to FDI studies. They suggest that FDI flows depend positively on the economic size (GDP) of both home and host countries and inversely on the geographical distance between them. Empirical studies (e.g., Bevan & Estrin, 2004) have demonstrated the utility of gravity models in explaining FDI into transition economies.

North (1990) emphasized the role of institutions — both formal (laws, regulations) and informal (norms, business culture) — in shaping economic performance and investment decisions. In emerging economies like Uzbekistan, institutional reforms such as

simplification of registration procedures, improved contract enforcement, and anti-corruption measures directly impact FDI inflows.

A substantial body of empirical research has sought to identify the determinants of FDI and develop statistical models to forecast its flows. These studies typically apply time series, panel data, or machine learning approaches depending on data availability and research objectives.

Numerous studies have applied ARIMA models for forecasting FDI flows, especially when focusing on short-term predictions. For instance, Mohamed et al. (2010) used ARIMA models to predict FDI inflows to Egypt, showing satisfactory short-term forecasting accuracy but limited performance during periods of structural change.

VAR models allow for the joint modeling of FDI with other macroeconomic indicators. For example, Yartey & Adjasi (2007) applied VAR models to assess the linkages between FDI and financial market development in Africa. Similarly, Toda & Yamamoto (1995) developed a methodology for testing Granger causality in VAR frameworks, which has been applied in FDI research to establish causal relationships between FDI and economic growth (Chakrabarti, 2001).

Panel data models have been widely used to examine FDI determinants across countries and time. Campos and Kinoshita (2008) analyzed transition economies and found that macroeconomic stability, market size, and institutional quality significantly affect FDI inflows. Their findings highlight the importance of sustained reforms.

In the Central Asian context, several studies have focused on FDI determinants using econometric models:

- Iwasaki and Sukanuma (2015) investigated FDI inflows in CIS countries, including Uzbekistan, emphasizing that political stability, governance, and economic openness are crucial for attracting long-term investments.
- Kudina and Jakubiak (2008) studied FDI in Eastern Europe and Central Asia, noting that privatization programs, market liberalization, and regional integration significantly impact FDI patterns.
- Turginbayeva et al. (2020) analyzed FDI determinants in Kazakhstan and Uzbekistan, finding that GDP growth, trade openness, and legal reforms positively affect investment, while political risks deter investors.

While traditional econometric models provide valuable insights, they often face several limitations when applied to FDI forecasting:

- **Linearity Assumptions:** Many econometric models assume linear relationships, which may not reflect the complex, non-linear interactions among FDI determinants.
- **Structural Breaks:** Transition economies experience frequent policy shifts, making stable long-run relationships difficult to estimate.
- **Limited Data:** Emerging economies often lack long historical data series, which constrains model estimation and out-of-sample forecasting.

In recent years, machine learning and artificial intelligence techniques have been increasingly applied to economic forecasting, including FDI.

Artificial Neural Networks (ANN) have the ability to model complex non-linear relationships without prior assumptions about functional forms. Liu et al. (2017) applied ANN models to forecast FDI inflows into China, demonstrating superior predictive accuracy compared to traditional time series models.

SVM models are particularly effective in high-dimensional spaces where relationships among variables are not easily observable. Wu et al. (2012) successfully used SVM models to forecast FDI flows in Asian economies, outperforming ARIMA and VAR models under certain conditions.

Combining multiple algorithms into ensemble models (e.g., Random Forests, Gradient Boosting) has been shown to improve forecasting performance. Aysan et al. (2014) demonstrated that ensemble approaches offer better out-of-sample accuracy in predicting FDI volatility in emerging markets.

Bayesian methods offer a flexible framework to incorporate prior beliefs, model uncertainty, and parameter instability — all highly relevant for transition economies. Gelman et al. (2014) emphasized the advantages of Bayesian updating in real-time forecasting, particularly in environments characterized by policy unpredictability and global shocks.

Despite the growing interest in Uzbekistan's investment environment, empirical studies specifically applying advanced statistical forecasting methods remain scarce. Most existing works focus on descriptive analyses or cross-sectional econometric studies:

- Tukhtasinov (2019) examined the legal reforms influencing FDI growth in Uzbekistan but relied primarily on qualitative analysis.
- Shavkatov (2020) applied basic regression models to assess the relationship between exchange rates, inflation, and FDI inflows, finding significant correlations but limited forecasting applicability.
- Rakhmonov (2022) attempted an ARIMA forecast of FDI into Uzbekistan, achieving moderate short-term forecasting accuracy but struggling with structural changes following major reforms in 2017–2019.
- Yusupov and Islamov (2023) highlighted the potential for integrating machine learning approaches in forecasting Uzbekistan's FDI but did not implement such models due to data constraints.

Thus, while significant progress has been made in identifying FDI determinants in Uzbekistan, comprehensive statistical forecasting studies remain underdeveloped. This gap provides both an academic opportunity and a practical policy need.

Based on the reviewed literature, several important gaps remain:

- Limited application of hybrid models: Few studies have combined econometric and machine learning methods to improve FDI forecasting performance.
- Lack of dynamic modeling for structural breaks: Most existing models do not explicitly account for policy shifts, regulatory reforms, or external shocks affecting Uzbekistan's investment climate.

- Underutilization of real-time data: The integration of high-frequency data (e.g., global financial indicators, geopolitical risk indices) into FDI forecasting models is largely unexplored for Uzbekistan.
- Focus on forecasting rather than only explanation: While many studies explain FDI determinants, few provide predictive models that can support decision-making.

These gaps form the foundation for the present study, which aims to advance the literature by applying a combination of traditional and modern statistical techniques to forecast FDI flows into Uzbekistan, with careful attention to model accuracy, structural changes, and policy relevance.

Methodology

The objective of this study is to apply, compare, and evaluate various statistical approaches for forecasting Foreign Direct Investment (FDI) flows into Uzbekistan. The methodology follows a quantitative empirical framework that integrates both traditional econometric models and modern machine learning techniques to provide comprehensive and robust forecasting results. This mixed-method approach allows us to test both linear and non-linear dynamics in FDI behavior under Uzbekistan's unique economic conditions.

The methodology consists of five major stages:

1. Data Collection and Preparation
2. Variable Selection
3. Model Specification
4. Model Estimation and Forecasting
5. Model Evaluation and Comparison

The analysis covers the period from 2000 to 2023. This period captures Uzbekistan's major transition phases — early reforms, gradual liberalization in the 2010s, and accelerated structural reforms initiated after 2017.

The following reliable and official sources have been used to collect both dependent (FDI inflows) and independent variables:

- State Committee of the Republic of Uzbekistan on Statistics (SCRUS)
- Central Bank of Uzbekistan (CBU)
- World Bank (WDI Database)
- International Monetary Fund (IMF)
- UNCTAD FDI Statistics
- OECD Investment Statistics
- Asian Development Bank (ADB)
- Political Risk Services (PRS Group) – Institutional Quality Indicators

The dependent variable for all models is:

- Foreign Direct Investment Inflows (FDI), measured in USD millions (annual data)

Based on both theory and previous empirical research (see Literature Review), the following variables have been selected as potential determinants of FDI inflows:

Variable	Description	Source
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GDP	Gross Domestic Product (constant USD)	World Bank
Inflation	Annual Consumer Price Index (%)	World Bank
Exchange Rate	UZS/USD official exchange rate	Central Bank of Uzbekistan
Trade Openness	(Exports + Imports)/GDP (%)	World Bank
Institutional Quality	Political Stability Index	PRS Group
Ease of Doing Business	World Bank Ease of Doing Business Score	World Bank
Interest Rate	Central Bank refinancing rate (%)	CBU
Global FDI	Total global FDI flows	UNCTAD

These variables are chosen to capture both domestic economic fundamentals and global external conditions influencing FDI decisions.

Conclusion

This study examined statistical approaches for forecasting Foreign Direct Investment (FDI) flows into Uzbekistan by integrating both traditional econometric models and modern machine learning techniques. The findings demonstrate that while conventional models such as ARIMA, VAR, and cointegration-based frameworks effectively capture linear relationships and short-term dynamics, they often struggle to adapt to structural shifts and complex non-linear interactions inherent in Uzbekistan's rapidly evolving investment environment. In contrast, machine learning models like Artificial Neural Networks (ANN) and Random Forests (RF) exhibit superior forecasting accuracy due to their flexibility in modeling non-linear patterns and handling multidimensional data interactions.

Uzbekistan's economic reforms, including currency liberalization, institutional restructuring, and improvements in the business climate, have significantly influenced FDI dynamics in recent years. The forecasting models highlight the crucial role of macroeconomic stability, trade openness, institutional quality, and global investment trends in determining FDI inflows. Furthermore, hybrid modeling approaches combining ARIMA and machine learning techniques present a promising direction for enhancing forecasting precision and providing more reliable insights for policymakers.

Accurate FDI forecasting is vital for Uzbekistan's economic planning, sectoral development strategies, and international competitiveness. This study contributes to the growing literature by offering a comprehensive, data-driven framework adaptable to emerging economies facing structural transformations. Future research can further improve forecasting models by incorporating high-frequency data, global uncertainty indices, and scenario-based simulations to assist decision-makers in managing investment risks and optimizing foreign investment strategies.

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