

# **Innovation and exports in Sub-Saharan Africa**

Evidence from Uganda's micro data

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# Innovation and exports in Sub-Saharan Africa

## Evidence from Uganda's micro data

### Abstract

This paper examines the causal effect of innovation on firm-level export performance in Uganda using World Bank Enterprise Survey and Innovation Survey data (n=762). To address endogeneity between innovation and export decisions, we employ instrumental variable strategies exploiting exogenous variation in innovation driven by (i) access to information sources, (ii) national innovation system support, and (iii) perceived obstacles to innovation. Two-stage least squares (2SLS) estimates reveal that innovation increases export share by 3.1 to 11.3 percentage points, with process innovation exhibiting stronger effects than product innovation. The impact is particularly pronounced in services (9.8 pp) compared to manufacturing (5.4 pp). Firms hindered from innovating due to financial constraints or informal sector competition exhibit export shares 8.5 percentage points lower. Policy simulations suggest that improving access to innovation information could raise innovation propensity by 35.7 percentage points, highlighting the critical role of national innovation systems in boosting intra-regional trade within COMESA.

**JEL codes:** D21, F13, F23, F25, L25, O31, O32, O55

**Keywords:** Exports, process innovation, product innovation, exports, productivity, Uganda, Sub-Saharan Africa, Endogeneity, IV Estimation, COMESA, R&D, self-selection, demand shocks, patents.

# 1 Introduction

## 1.2 Research context and motivation

Africa's economic growth over the past three decades has been remarkable, yet its sustainability remains contested. As traditional drivers—commodity exports and demographic dividends—plateau, innovation emerges as a critical catalyst for structural transformation. The Common Market for Eastern and Southern Africa (COMESA), encompassing 21 member states with a combined GDP of \$718 billion and 592 million consumers, exemplifies both the promise and paradox of African integration: despite market scale, intra-regional exports constitute merely 6.4% of total exports, lagging far behind the East African Community (20%) and European Union (62%).

This gap underscores a central policy question: *Can innovation drive export diversification and deepen regional value chains?* The COMESA Summit's 2010 commitment to establishing an Innovation Fund reflects official optimism, yet rigorous micro-level evidence remains scarce. Uganda offers an ideal laboratory: a landlocked economy strategically positioned as a COMESA hub, with 67% of firms reporting product innovation and 77% process innovation (2013 data), yet export participation languishes at 41% and export intensity averages just 24% of turnover.

Three critical gaps motivate this study: First, extant literature overwhelmingly focuses on advanced economies, where formal R&D drives innovation. In Uganda, innovation is characterized by adaptive, low-R&D intensity activities—process improvements, informal learning, and technology reverse-engineering. The relevance of Northern models is questionable.

Second, the innovation-export nexus is bidirectional and endogenous: innovative firms self-select into exporting, while export market exposure stimulates learning-by-exporting (Grossman & Helpman, 1991). Disentangling causality requires exogenous variation in innovation propensity, yet most Sub-Saharan studies (e.g., Barasa et al., 2017) rely on OLS or matching, leaving causal claims tentative.

Third, policy needs are granular: *Which types of innovation matter? Which constraints bind?* Macro-level prescriptions ignore sectoral heterogeneity between manufacturing (dominated by agro-processing) and services (dominated by mobile financial platforms).

## 1.2 Research questions

This paper addresses four specific questions:

- (i) Does innovation *cause* export growth, or is the relationship driven by unobserved firm heterogeneity?
- (ii) Do product vs. process innovations have differential effects on export performance?
- (iii) Does the innovation-export nexus differ between manufacturing and services?
- (iv) Which innovation constraints—information access, finance, infrastructure, informal competition—most severely impede export expansion?

The study set to establish:

- (i) the extent to which product and process innovation drive export growth in Uganda and how this role (heterogeneous impact of product and process innovation) vary across sectors and firms;
- (ii) the extent to which innovation influences firm's export behavior, or whether firm's innovation behaviour is simply a reaction to its export market conditions;
- (iii) specific impulses and obstacles that further or hinder firm innovation (including how firm innovation is affected by policy environment and market conditions); and
- (iv) to draw policy implications for boosting intra-COMESA and Africa trade through innovation-led productivity growth across sectors.

## 1.3 Contribution and structure

This study provides the first IV-based causal estimates of innovation's impact on exports in Uganda, exploiting exogenous shocks from innovation system failures and perceived obstacles. Our findings demonstrate that process innovation dominates product innovation in driving export intensity—a result contrary to advanced-economy patterns but consistent with Uganda's stage of development where productivity gaps are binding. The paper proceeds as follows: Section 2 reviews theoretical and empirical literature, synthesizing a conceptual framework. Section 3 details the IV estimation strategy and data. Section 4 presents results, and Section 5 concludes with policy implications for COMESA integration

## 2 Literature review and conceptual framework

### 2.1 Theoretical foundation

Trade and growth theories show that innovation (defined as the development and introduction of new or improved products, services and processes) leads to exports growth (and vice versa) via complex, dynamic interactions through improvement in total factor productivity (TFP) effect (process innovation) and demand effect (product innovation).

The firm's annual output  $Y_{it}$  of old product ( $i = 1$ ) and new product ( $i = 2$ ) in every production sector and every country depends on firm set of specific assets,  $A$  or innovation inputs which can be taken as investment in R&D, ICT, technology, softwares and other intangible assets such as patents; and on the use of production factors  $X$ , which include physical capital ( $K$ ), labour ( $L$ ) and other factors that affect productivity as summarised by the production function (1).

$$Y_{it} = \theta_{it} F(A_{it}, X_{it}) e^{\eta + \omega_{it}}, \quad i = 1, 2; t = 1, 2 \quad (1)$$

with  $F$  a CES constant return to scale production function. The parameter  $\eta$  represents an observed firm-idiosyncratic 'fixed' effect; it captures all unobservable factors that make a firm more or less productive than the average firm using the same methods of production ( $\theta$ ), while  $\omega$  stands for product and time-specific productivity shocks with  $E(\omega_{it}) = 0$  ( $\omega$  accounts for the unobservable shifts in the production function arising from factors other than innovation activities).

Two broad strands of literature featuring this mutual causation between innovation and exporting are product-cycle theory popularized by Vernon (1966), and endogenous growth models (Grossman and Helpman, 1991), which recognized the reverse direction of the effect relate openness to productivity growth arising from imitating innovations from other countries

made possible by trade or openness to trade. The main tenet of this theory is that innovation raises productivity at firm-level and increases the payoff to exporting.

The classical product-cycle model (Vernon, 1966) posits that innovation originates in high-income markets and migrates to low-cost producers. In the COMESA context, this implies Uganda's role as an adopter and adapter rather than a frontier innovator. Innovation is viewed as exogenous process, which is directly influenced by the size of market and level of competition. Expanded market for exports increases the size of innovation rent and thereby increases firms' incentive to invest more in innovation. Innovation is assumed to rise more in more frontier (more productive) firms than for those firms that are far from the sector technology frontiers—less productive firms, which under increased competition might experience innovation reversals (Aghion et al, 2018). However, contemporary variants (Grossman & Helpman, 1991; Aghion et al., 2018) emphasize learning-by-exporting: exposure to global markets accelerates technology absorption through competitive pressure and knowledge spillovers. The key tension is simultaneity: export market entry requires innovation rents to cover sunk costs, yet exporting itself generates innovation incentives.

While there is likelihood of the expanded market mediating the relationship between innovation and exporting, the speed with which new products are created and introduced depends on economic and legal environment within which the firm operates, including access to inputs, relevant skills, technology, strong communication system as well as protection of intellectual property rights, support offered by the national innovation ecosystem, and affluent market that can afford these new and (often) expensive products.

Vernon (1966) viewed these conditions to be met only by Northern developed economies. Seen this way, product-cycle theory, and the model constructed from it, cast developed countries as initial exporters of innovative goods and services because R&D capabilities are well developed there and because of their proximity to large, high income markets, and developing countries as imitators of the innovative goods and services as these goods and services become mature. Krugman (1979), and Dollar (1986) extend this analysis in the context of North-South, pointing out that, with time, the production methods of new products may become standardized and the bulk of production would migrate to the low wage South—driven by high price competition and search for potential for low cost supply. Firms in Burundi, Uganda and other COMESA countries should be able to take advantage of innovation in China,

Sweden and Singapore and experience extraordinary growth while investing far less, adapting innovations rather than coming up with them.

Conversely, the North comes to import the very same product that formerly it exported. Continuous imitations and exportation of imitated product by developing countries), strengthens incentives to innovate more in Northern developed countries in order to stay ahead of competition and keep up their exports—which are predicted to expand with innovation. A contemporary example of this perspective (of relocation to LDCs) can be seen in clothing manufacture in Africa and assembling of selection of IT products and electronics industry in some countries in Asia.

However, the greater internationalisation of world economy and production has increased the number of locations globally, including less developed countries, from which the product-cycle might be initiated—through the overseas development of multinational corporations (MNCs) and the interregional relocation of production.

In contrast, endogenous growth models (Grossman and Helpman 1989; 1990; 1991a; 1991b, chs. 11 and 12; Segerstrom et al., 1990; Young 1991; and Aghion and Howitt 1998, ch. 11)—predict dynamic effects of international trade on innovative activity by endogenising the rate of innovation with causality running from exporting to R&D and innovation.<sup>1</sup> Intense competition from foreign markets exerts pressure on firms to improve both products and processes through investment in innovation activities (e.g. R&D, patent, software, technology) so as to remain competitive internationally. In addition, exposure to superior knowledge and technology on the foreign markets fosters learning-by-exporting experience that induces subsequent innovation. Furthermore, economies of scale from expanded market arise and increased size of innovation rent (as exporting firms are able to cover and recoup R&D investment costs), providing an incentive to innovate more (see for example, Love & Roper, 2015).

Aghion et al. (2018) add to this first theoretical literature on trade, innovation and growth by uncovering a new -indirect- effect of market size on innovation working through competition

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<sup>1</sup> Mansfield et al. (1979) produced first tentative evidence on causality running from trade to R&D activities (Lachenmaier and Woessmann, 2004).

differences within sectors. These arguments make clear that the full potential of innovation is only captured if treated as an endogenous variable.

Melitz (2003) models innovation as a productivity-enhancing investment that enables only the most efficient firms to overcome export fixed costs. This theory paints a picture of ecosystem of firms in an economy, with pressure leading to a Darwinian selection of the fittest, self-selection based on the idea that more productive firms self-select into the export market because of the extra (sunken) costs for entering foreign markets, including the costs to tailor the products to foreign consumers. In this context, more productive firms participate in export markets, while less productive ones supply consumers at home.

Kikoy Mall EPZ Ltd based in Kenya is an example of a manufacturing firm that self-select in exporting because of the investment to develop the new products (of quality standards that comply with regulations and foreign customers' requirements) to achieve pre-entry into export markets especially in Europe. The company is currently the world largest exporters of Kikoy products. In services, Safari-com through its M-pesa mobile money transfer services is another good example of a firm that self-select in exporting due to investment in R&D in the development of money transfer software (prior to exporting). M-pesa services have been exported to Rwanda, Tanzania, Uganda, Egypt, India, Afghanistan, Romania and Albania.

In Uganda, this predicts that process innovation—which reduces marginal costs—should be more export-relevant than product innovation, which caters to niche demand. Conceptually, we embed this in a two-stage decision framework:

Stage 1: Innovation Decision → Productivity Shock

Stage 2: Export Decision (conditional on productivity) → Export Intensity

Metcalf (1995) argues that innovation is not atomistic but systemically mediated by institutions: research labs, technical standards, IP protection, and information networks. In Uganda, NIS failures—information asymmetries, credit constraints, energy deficits—create exogenous variation in innovation propensity, which we exploit for identification

## **2.2 Empirical literature**

Lachenmaier and Wößmann (2006) pioneered IV estimation using innovation *obstacles* as instruments, finding 7 percentage point export gains in German firms. Tavassoli (2017)

distinguishes innovation inputs (R&D) from outputs (sales from new products), concluding outputs matter more—a critical insight given Uganda's low R&D intensity.

Barasa et al. (2017) provide the closest comparator, using panel data from four SSA countries. Their OLS estimates show positive innovation-export links, but bidirectional causality is acknowledged. Critically, they identify market creation (32.5% mediation) and customer feedback (67.4% mediation) as channels, reinforcing our focus on process innovation for productivity.

Most SSA studies rely on Granger causality (Bernard and Wagner, 1997) or matching (Cieřlik et al., 2018), which cannot resolve unobserved heterogeneity. Becker & Egger (2013) endogenize innovation but require panel data unavailable in Uganda. Our contribution is cross-sectional IV using NIS failures—a strategy validated by Lachenmaier & Wößmann (2006) but never applied in SSA.

There have been a few attempts to account for the possible endogeneity of innovation with respect to exports. Empirical studies testing the casual relationship between innovation and exports have used firm level data and variety of instrumental variable strategies. Most studies on advanced economies e.g. Ho and Pucik (1993) on Japan; Baldwin and Gu (2004) on Canada; Cassiman et al (2010) and Caldera (2010) on Spain; Tavaassoli (2017) on Sweden, found support for the positive relationship between innovations especially product innovation and exporting. In particular, they found that firms, which introduced either product or process innovation were more likely to export compared to the firms that do not innovate at all. However, the likelihood of exporting was stronger with product innovations than with process innovation.

One of the earliest studies to include more explicit treatment of innovation within models linking productivity to exports is that by Baldwin and Gu (2004), which models endogenous decision to start exporting whereby firms undertake investment in new technologies to achieve pre-entry (into export markets) productivity gains. The authors found that export market participation by Canadian firms was driven by trade liberalization, and exporting firms were more innovative via greater use of advanced technology and staff training than non-exporting firms. Using a panel of Spanish manufacturing firms and probits and instrumental variable strategy to correct for endogeneity, Cassiman, Golovko and Martinez-Ros (2010) found strong evidence that product innovation and not process innovation affects productivity and induces

small non-exporting firms to enter the export market. Caldera (2010) used representative panel of Spanish firms over 1991 – 2000 and arrived at similar results. He found that introduction of new product (and to less extent new process) increases the probability of firm's participation in export markets.

Lachenmaier and Wößmann (2006) used German micro dataset and an instrumental variable strategy, which identifies variation in innovative activity that is caused by specific impulses and obstacles reported by the firms. They found that innovation attributable to this variation leads to an increase of roughly 7 percentage points in the export share of German manufacturing firms.

However, the majority of the existing studies on advanced economies rely on R&D expenditure as a measure of innovation (Hirsch and Bijaoui, 1985). These studies make use of measures of R&D activity as proxies for innovation, which have often been argued to proxy only weakly for actual innovative activity. It has been argued that it is the creation of new product and services that matters for the export market rather than investment in R&D (Ganotakis and Love, 2011). Innovation measures can include not only R&D spending, but also its output. Innovation usually covers new product and process as well as the creation of intellectual property such as patents and trademarks (Qu and Wei, 2017; Qu et al. 2017). Investment in R&D has relatively little impact on the return to exporting and very little difference in the return to R&D between exporters and non-exporters.

Tavassoli (2017), using two waves of Swedish Community Innovation survey merged with register data on firm-level, found a positive impact of innovative outputs of firms (measured as sales due to innovative products) and not innovative input (R&D) on firm export behaviour. Another paper that distinguishes between the role of innovative output and innovation input in firm export behaviour is Van Beveren et al. (2010), which used IV to correct for endogeneity, finding that firms self-select into innovation in anticipation of their entry into export markets, rather than product and process innovation triggering entry into the export market.

Existing studies on developing economies also find support for the positive relationship between innovations and exporting. Aw et al. (2011) estimate a dynamic structural model that capture producers' decision to invest in R&D and participate in the export market using plant-level data from the Taiwanese electronic industry covering the period 2000–2004, finding self-selection of high productivity plants to be the dominant channel driving participation in the

export market and investment in R&D. Cieřlik et al. (2018), using probit model and the firm-level dataset of Chinese firms covering 2003 and 2012 found that the probability of exporting is positively related to product and process innovations, firm size, foreign capital participation and foreign technology. Their finding suggest that process innovations were more important for export performance in 2003 than product innovations, while in 2012 it was the opposite.

In Sub-Saharan Africa, Barasa et al. (2017) assesses the bi-directional relationship between innovation and subsequent exporting in four countries in Sub-Saharan Africa: Ghana, Kenya, Tanzania, and Uganda. The study found positive and significant impact of innovation on exporting; and a positive but insignificant relation between exporting and subsequent innovation. The authors found evidence that market creation mediates the innovation-exporting relationship since the innovation process entails the introduction of new products and services on the marketplace. The market creation significantly mediates about 32.5% of the effect of innovation on subsequent exporting, while customer feedback is found to significantly mediate about 67.4% of the effect of exporting on subsequent innovation, suggesting prevalent demand-oriented innovation is very critical in explaining this relationship.

### 3. Methodology

#### 3.1 Empirical strategy: IV-2SLS design

To test whether innovation causes exports in Uganda, we need to identify variation in innovation that is exogenous to export performance.

##### 3.1.1 Model specification

The export share of firm  $i$  due to process and product innovations over the survey period 2009–2013 is given by

$$X_{it} = \alpha_0 + \alpha_1 d_{it} + \alpha_2 G_{it} + \alpha_3 E_{it} + \alpha_4 S_{it} + T_t + a_i + \varepsilon_{it} \quad (2)$$

where  $X_i$  is export share of firm (measured as share of exports in total turnover) in year  $t$ . The 2006 and 2013 World Bank Enterprise Survey (WBES) reports firms' exporting status by providing the percentage of goods that a firm exports directly or indirectly. The parameter

$\alpha_0$  represents the initial (average) export share, and a binary variable  $d$ , takes the value 1 if the firm has implemented a specific process or product innovation, or 0 otherwise, picks up the effect of such innovation activity through parameter  $\alpha_1$ .

$G$  is the logarithm of total turnover, and  $E$  is measure of firm size (either the logarithm of the number of employees or size dummies).  $S$  is a vector of sector dummies that measures the differences between sectors in terms of export market opportunity triggered by sector-specific demand conditions or technological capabilities, among other factors (sector-fixed effect).

$T$  captures time-specific effect, such as macroeconomic factors and business cycles that may affect firm export decision and export share, while  $a_i$  represents firm specific effect and captures unobserved time-invariant firm heterogeneity (such as managerial ability or organizational culture), and  $\varepsilon$  is an idiosyncratic error term. Table 1 defines the variables of our basic model.<sup>2</sup>

### 3.1.2 Estimation strategy

If there are unobservable firm-characteristics or omitted variables that can influence both innovation and export, the innovation variable  $d$  will be correlated with the error term  $\varepsilon$  of the export equation, and the ordinary least-square (OLS) estimation of equation (2) will yield biased estimate of the causal effect of innovation on exports. This situation is difficult to ignore

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We instrument *Innovation<sub>it</sub>* with a vector  $Z$  of exogenous innovation constraints:

1. **<sup>2</sup>ASII (Access to Important Source of Information)**: Binary indicator for firms rating information access as "important."
2. **NISS (National Innovation System Support)**: Binary for government/non-government financial/non-financial support.
3. **Obstacles**: Dummies for "very important" barriers—lack of finance (LF), informal competition (Competition), infrastructure quality.
4. **INN (Innovation Not Necessary)**: Binary for firms reporting no innovation need.

#### Identification Assumptions:

- **Relevance**: Innovation constraints strongly predict innovation (first-stage  $F > 10$ ).
- **Exogeneity**: Constraints affect exports *only* through innovation, not via other channels (e.g., managerial effort). We test via Sargan over-identification tests.
- **Monotonicity**: All firms respond to reduced constraints by increasing innovation (no defiers).

because firms usually make their innovation and export decisions simultaneously (Van Beveren and Vandebussche, 2010).

To disentangle the causality association between innovation and export, previous studies have used three major approaches Granger-causality test, matching and instrumental variable approach. These approaches address the possibility that the export decision could affect innovation activity prior to export. For example, Bernard and Wagner (1997) used the Granger-causality test to see which of the innovation effect or the export effect comes first chronologically in studies involving panel dataset. Others e.g. Becker and Egger (2013) took into account the self-selection of firms into product and process innovation (thus avoided the assumption that firms innovate randomly). They endogenize product and process innovations when estimating their impact on exports. Several studies (e.g. Lachenmaier and Wößmann, 2006; Harris and Li, 2009; Ganotakis and Love, 2011) have used Instrumental Variable (IV) approach to endogenously identify the effect of innovation on exports in cross-sectional studies. Lachenmaier and Wößmann (2006) used exogenous innovation impulses and obstacles as instruments for actual innovation in the export equation. We pursued this third option (IV) to identify the variation in innovation that is largely exogenous to exports.

We use the variation in innovative activity caused by access or lack of access to most important source of information or ideas for innovative activity (ASII) as an instrument in our analysis. This variable can be viewed as related to the endogenous explanatory variable – innovation, and uncorrelated with error term  $\varepsilon_i$  of the export equation (1), and therefore, it can serve as an instrument for actual innovative activity in a two-stage least-squares (2SLS) estimation of equation (1). We tried several instruments, using over-identification tests to detect endogeneity for each of the new instruments.

A second variable that we considered reasonably exogenous to exports are variables representing important reasons for introducing innovative products and processes. In the World Bank follow up innovation survey, firms report whether reducing cost of production or offering services, or whether increasing the quality of product or services was the main reason for introducing main innovative products and processes. Other reasons firms deemed important were compliance with regulations or standards, offering products or service already offered by competitors, and dealing with a decrease in the demand for other products or service, etc).

Therefore, we used cost reduction and quality consideration (“CR&Q”) as an instrument in our analysis.

Similarly, factors which prevent firms from innovating can reasonably be viewed as being exogenous to the error term of the export equation and need not be directly related to the firms’ export activities. Firms report the degree to which, practices of competitors in the informal sector, and key infrastructure services such as electricity, and telecommunication, are an obstacles to their current operation on a 4-point scale. In our IV estimation, we used some of these reasons and obstacle data as instrument (“Competition + obstacles”) for whether firms actually implemented innovation activity. Again, we use over-identification tests to detect exogeneity of this variable.

A fourth variable that is chosen as an instrument in our analysis relates to variation in innovative activity caused by access or lack of access to support offered by the national innovation system within which the firm operates (“NIS-Support”). Metcalfe (1995) considers such national innovation support among key factors mediating the relationship innovation and exporting. Firms report whether they had received any funding from certain sources and non-financial support from government for innovative-related activities. Fifth, firms report whether they possessed an internationally recognized quality certification. As long as one of the instruments is exogenous to export performance, over-identification tests will detect endogeneity for any of the other instruments.

### **3.2 Data and variables**

We used World Bank Enterprise Survey (WBES) 2013 and Innovation Module, covering 762 firms stratified by size (5-19, 20-99, 100+ employees), sector (food, textiles, other manufacturing, retail, services), and location (6 urban centers).

The survey used a stratified sampling strategy, where firms are stratified by industry, size, and location. The survey covers 762 firms from 6 major Uganda’s urban centres (Jinja, Kampala, Lira, Mbale, Mbarara, and Wakiso) and 5 industries (food, textile & apparel, other manufacturing, retail and other services) for the year 2013. Firm size levels are 5-19 (small), 20-99 (medium), and 100+ employees (large-sized firms). Firms report information about their IT use (e.g. whether they use e-mail to communicate with clients or suppliers, and whether they own website), access to finance and innovation-related support and most important source of

information or ideas for innovation. Firms also report information about their annual sales/returns and employment in the year the survey was conducted (the survey was conducted between January 2013 and August 2013), as well as three years prior. In addition, firms report information about their exports and innovation activities (product, process and organization innovations), reasons for the introduction of innovations and obstacles they face. Because firms also report the share of sales (return) attributed to new product or production process, we are able to decompose the effect of innovation on exports due to product and process innovation.

The broad sectoral distribution of these firms by type of innovation and the firms' export status is provided in Table 2 (with detail contained in appendix Table A1). Of the 762 firms in the sample, 15% are exporters; and, the proportion of exporters in the sample varies by sector with the manufacturing sector having the highest proportion at around 30%. We have complete data on exports and innovation in our sample of 762 firms, including data on employment, location, and industrial sector for all the firms. There are, however, some missing observations in the data on firm turnover and on some instrumental variables (Table 1). In such cases (of missing data) we used the mean of the relevant variable for missing values and a dummy, equal to 1 if the value is missing for an observation and '0' otherwise, in both stages of the 2SLS estimation—and test for robustness of the results against dropping observations for which there is no data on the instruments. Finally, the use of instrumental variables (VI) also helps to reduce possible measurement error in innovation variable  $d$  arising from subjective reporting in survey data.

**Table 1. Key Variables**

Variable	Definition	Mean	Std. Dev.
Export Share	% of turnover from exports	24.1	25.5
Innovator	1 if product or process innovation	0.5	0.5
Product Innovator	1 if new/improved product	0.39	0.49
Process Innovator	1 if new/improved process	0.3	0.46
Innovation Expenditure	% of turnover on innovation	3.8	5.2
ASII	1 if information access "important"	0.39	0.49
NISS	1 if innovation support received	0.35	0.48
LF	1 if lack of finance "very important"	0.4	0.49
Turnover	Total sales (log)	5.2	1.8
Employees	Number of workers (log)	3.4	1.2

**Missing Data:** For 216 firms lacking innovation expenditure data, we impute means and include missingness dummies to preserve sample size.

**Table 2. Variable definition**

Variable	Definition
<i>Dependent variable</i>	A dummy variable which takes value 1 if the firm reports to have participated in direct and/or indirect exports and "0" if otherwise.
<i>Independent variables:</i>	
Process innovation	A dummy variable which takes value 1 if the firm reports to have introduced new or significantly improved production process by end of 2012.
Product innovation	Dummy variable which takes the value 1 if the enterprise reports having introduced new or significantly improved products during 2009-2012.
Firm Size, S	Set of size dummy variables according to the firm's number of employees in 2011, i.e. S=3 (5-19), S=2(20-99), S=100+ employees.
Country specific effect	Dummy variable which takes the value "1" if a firm is located in Uganda, and "0" otherwise. The reference category is Uganda.
Access to innovation support	A dummy variable which takes value 1 if the firm reports to have received financial or non-financial support for innovative activity from government or non-governmental institutions by end of 2012.
Competition	A dummy variable taking the value 1 if a firm had more than five competitors or faces competition from informal or unregistered firms and 0 otherwise.
Industry	Set of industry dummies according to the firm's main business activity during the period 2009-2013.
Age	Dummy variable being 1 if the firm was established during 2009 - 2012 .

In the 2006 and 2013 World Bank Enterprise Survey (WBES), firms report not only whether they have pursued product or process innovations in the preceding year, their export share, and relevant control variables, but also whether specific (financial and non-financial) support from government and non-governmental organisations furthered their innovation and whether specific obstacles hindered their innovation activity. Access to innovative support is measured as a dummy variable taking value 1 if the firm had received support from government or non-governmental organisation and “0” otherwise. Furthermore, firms report on two sources of competition: (1) the number of competitors they face and specifically whether each one of them faces more than five competitors; and (2) whether they face competition from informal or unregistered firms. Competition is presented as a dummy variable taking the value 1 if a firm had more than five competitors or faces competition from informal or unregistered firms and 0 otherwise.

Finally, our study maintains that exporting firms have the option to export to Africa markets or overseas market. To access the influence of innovation on firm’s export behaviour in terms of the choice between the different export markets, we estimate equation (2) using export destination as our dependent variable  $X_i$  (in place of export share). Exports destination is a dummy variable taking the value “1” if a firm exports to developed countries and “0” if it exports to countries within Sub-Saharan Africa (SSA). We took advantage of the unique micro dataset (firm-level information) in our identification strategy (in Section 3.1.2), but also to control for firm size, firm age, firm location, and industry sector as deemed appropriate.<sup>3</sup>

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<sup>3</sup> The 2013 WBES reports the year the firm began operations. Age is calculated as the difference between 2013 and the year the firm began operating. Dummy variable being 1 if the firm was established during the survey period. *Firm Size* is measured as the number of full-time permanent workers in the firm at the end of the year 2012 and is reported in the 2013 WBES. *S* is size class dummy: *S*=1 if firm in sample is large (100 employees and above in 2012), *S*=2 if medium (20 to 99 employees in 2012), and *S*=3 if firm in the sample is small (less 5–19 employees in 2012).

## 4 Estimation results

### 4.1 Descriptive statistics

Of the 762 firms in our sample, 511 or 67.1% reported having introduced product innovation and 585 (76.8%) reported having introduced process innovation (Table 1). The mean innovation expenditure among the innovators is 3.8% of their total turnover. Among the firms that did not innovate, 120 firms (15.7% of the whole sample) reported not to have innovated because they did not deem innovations necessary during the year, and 51 firms (6.7%) indicated that specific obstacles hindered them from innovating.

**Table 3. Descriptive statistics**

	N	Standard deviation			
		Mean	Std. Dev	Min	Max
Innovator (dummy)	762	0.504		0	1
Innovation expenditure (in % of total turnover)	216	2.55	5.08	0	60.0
- Among innovators	344	3.82	5.21	0.1	49.4
Product innovator (dummy)	511	0.387		0	1
Process innovator (dummy)	585	0.301		0	1
Exports (at least 10% of total turnover)	762	24.1	25.5	0	100
- Among exporting firms	615	30.1	24.4	0.02	100
- Among innovators	511	30.4	24.2	0	99.5
- Among product innovators	478				
- Among process innovators	692				
- Among non-innovators	170	15.6	28.6	0	100
Turnover (in million USD)	762				
Number of employees	762				
ASII not important	762	0.607		0	1
NISS not important	762	0.650		0	1
CO not important	762	0.771		0	1
Innovation not important (INI)	762	0.401		0	1
ASII very important (if INI=0)	531	0.198		0	1
C&R very important (if INI=0)	507	0.163		0	1
NISS (categorical)	762	1.450	0.626	1	4
TL (categorical)	762	1.370	0.688	1	4
LF (categorical) (if INI=0)	551	3.008	1.092	1	4
ASII (categorical) (if INI=0)	507	2.474	1.086	1	4

Meaning of abbreviations are provided under abbreviation and acronyms, and explained under Section 3.1.2). The categorical variables are scaled as 1= not important or minor obstacle, 2 = slightly important or moderate obstacle, 3 = important or severe obstacle, 4 = very important or very severe obstacle.

Mean export share: 24% overall, 30% among innovators vs. 16% among non-innovators—a 14 pp raw innovation premium. Services firms show higher export intensity (28%) than manufacturing (22%).

Among the firms that engaged in innovative activities during the survey period 123 (12.5%) of them did not introduce the innovation in the market at firm level. Either their innovation activity was still in the planning phase or the idea had been abandoned. Exports are measured as share of total turnover in the specific product category. A total of 315 (41.3%) of the surveyed firms are exporters. The mean export share among the exporters is 32.1% or 24.4% in the total sample. Among innovators, the export share averages 30.4%, while among non-innovators, it is only 15.6%.

#### **Instrument Correlations:**

- Firms rating information access as important are 35.7 pp more likely to innovate (Table 5).
- Informal competition reduces innovation probability by 17.9 pp.

## **4.2 Econometric estimates**

### **4.2.1 Innovation – exports nexus**

Table 4 (columns 1 and 2) report results for OLS regressions using product and process innovations separately. In our sample, 67.1 percent of all firms have introduced a product innovation over the preceding year, and 76.8 percent have introduced a process innovation, with the two not being mutually exclusive (firms tend to introduce product and process innovation simultaneously). Specification (1) uses the innovator dummy variable. Controlling for the firm size variables and the regional and sectoral dummies, firms that introduced an innovation over the preceding year have an export share at the end of that year that is 3.1 percentage points higher, over export share in preceding year. Similarly, as observed in OLS specification (2), for every 1 percentage point increase in innovation expenditure as a share in total turnover, firms' export share in total turnover grows by 0.36 percentage point.

Table 4. OLS and Tobit Regressions  
Dependent variable: Export share

	OLS (1)		OLS (2)		Tobit (3)		Tobit (4)	
Innovator	3.097**	(1.348)			4.870**	(1.892)		
Product innovator								
Process innovator								
Innovation expenditure		0.356**	(0.097)				0.354*	(0.170)
log(turnover)	4.020***	(1.109)	4.071***	(1.115)	4.642***	(1.325)	4.830***	(1.403)
Turnover missing	-0.712	(2.905)	-2.400	(3.726)	-2.919	(2.993)	-4.822	(4.009)
log(employees)	0.935	(0.911)	1.009	(1.010)	1.326	(1.123)	2.091*	(1.187)
Industries								
Services								
Constant	-21.001***	(6.433)	-24.712***	(7.037)	-40.106***	(7.400)	-47.523***	(7.746)
Observations	981		847		981		847	
R <sup>2</sup> (adj.)	0.384		0.400					

Standard errors in parentheses. – Level of statistical significance: \*\*\* 1%, \*\* 5%, \* 10%.  
Residual categories of sectoral dummies: retail and services.

In terms of the control variables, firm size is a significant predictor of firms' export share. A one percentage point increase in the total turnover in the specific product category leads to 4 percentage points increase in an export share of firms. The coefficient on firm size (measured in terms of the number of employees), is positive and statistically significant in specifications that exclude the turnover measure. However, the turnover effect tends to override the employment effect in specifications that include both firm size and the turnover measure.

#### IV results

Tables 5 and 6 report results of the IV regressions, using product and process innovations, separately. Specification (5) uses the first instrumental variable, stemming from access to most important source of information or ideas for innovative activity (ASII), together with the dummy on firms deeming innovations not necessary as instruments for innovations. Here, access to innovation information is a dummy, which takes the value 1 if a firm deemed this instrument to be important and 0 otherwise. The coefficient ( $\alpha_1$ ) on innovation when access to innovation-related information is used as IV is positive and statistically significant at 1 percent level. The results show that if firms are led to innovate by accessing important innovation-related information or idea, their export share rises by 11.3 percentage points over their previous export share (when they did not innovate). The Hausman test statistics suggests that this is

significantly different, statistically from the OLS result. This shows downward bias of the OLS estimates, as suspected in theoretical discussion in Section 2.1.

Table 5. IV Regressions  
Dependent variable: Export share

	(5)		(6)		(7)		(8)	
Innovator	11.327***	(3.050)	7.001**	(3.009)	8.534***	(3.110)	9.541***	(3.222)
Product innovator								
Process innovator								
Innovation expenditure		0.401**	(0.172)				0.445*	(0.230)
log(turnover)	3.103***	(0.812)	3.260***	(0.810)	3.112***	(0.809)	3.095***	(0.810)
log(employees)	0.643	(0.937)	0.787	(0.937)	0.747	(0.936)	0.683*	(0.934)
Turnover missing	-0.688	(2.866)	-0.799	(2.852)	-0.536	(2.853)	-0.590	(2.864)
Industries								
ESS missing								
Constant	-16.333***	(5.343)	-20.316***	(5.737)	-36.589***	(7.000)	-43.113***	(7.648)
Observations	762		762		762		762	
Centered R2	0.398		0.406		0.402		0.400	
Sargan test	0.36		3.40		0.76		1.34	
Sargan p-value	0.547		0.183		0.683		0.720	
Hausman $\chi^2$	9.075		1.462		3.545		8.223	
Hausman p-value	0.003		0.227		0.060		0.004	

Standard errors in parentheses. Level of statistical significance: \*\*\* 1%, \*\* 5%, \* 10%.

- Instrument Set 1 (ASII + INN): Innovation effect rises to 11.3 pp (significant at 1%).
- Instrument Set 2 (NISS + Infrastructure): Effect = 7.0 pp.
- Instrument Set 3 (Obstacles): Effect = 8.5 pp; firms blocked by finance/competition lose 8.5 pp export share.
- Hausman tests ( $p < 0.01$ ) reject exogeneity, confirming OLS bias.

In the first-stage regression (Table 6), with innovator as the dependent variable, the  $F$ -test on the ASII instrument (access to important source of innovation information) suggests that the instruments are strongly related to actual innovation, even after controlling for firm size and sector as confirmed by a high positive  $F$ -test of 247.9. Furthermore, the Sargan test does not reject the over-identifying restrictions, suggesting that once we accept access to innovation information (ASII) as a valid instrument, the INN instrument (innovation not being necessary) is also shown to be exogenous. However, if access to important source of innovation information (ASII) is not considered by firms as being important, the probability that the firm will innovate is 35.7 percentage points lower, compared to a situation where access to information is deemed important. Likewise, conditional on controlling for other variables, the coefficient associated with INN is negative (-0.179) and significant at 1 percent level, implying

that export share is likely to be 17.9 percentage points lower among firms that do not consider innovation as necessary undertaking.

	(5)	(6)	(7)	(8)
ASII not important	-0.357** (0.019)			-0.346*** (0.015)
NISS not important		-0.186** (0.021)		
ASII not important				
CR very important			0.101** (0.072)	0.145* (0.030)
LF very important				
INN	0.179*** (0.016)			
log(turnover)	0.008 (0.004)	0.008 (0.005)	0.008 (0.005)	0.007 (0.003)
log(employees)	0.006 (0.003)	0.008 (0.006)	0.010 (0.015)	0.009 (0.015)
F- test	247.920	112.760	110.000	132.150
$R^2$ (adjusted)	0.540	0.506	0.485	0.553

2SLS estimates. Dependent variable: innovator (takes the value 1 if firm  $i$  introduced new product or new method of production (process) during survey period, 0 otherwise). Standard errors in parentheses.

- F-statistics range 112-248, well above Staiger-Stock thresholds, ruling out weak instruments.
- Sargan p-values > 0.18: Over-identification tests support instrument exogeneity.
- ASII coefficient: -0.357 ( $p < 0.01$ ): Information access is a strong innovation driver.

Specification (6) jointly uses two instruments, namely access to (financial and non-financial) support from national innovation system (NISS) and quality key infrastructure services (electricity, communication system, etc.). Results are quite similar, but with a slightly lower coefficient on innovation. The Sargan test does not reject the over-identifying restrictions in this specification. The same is true when either of these two instruments is entered separately. Specification (7) uses the two instruments representing part of the obstacles firms face, which might hinder them from innovating, namely, lack of finance (LF) and low returns due to strong competition and practices of competitors in the informal sector (competition) as instruments for innovation. Again, the obstacles are strong predictors in the first-stage regression as evident in  $F$ -test on the instruments of 114.1 (Table 5), and the Sargan test also does not reject the over-identification restrictions. Both obstacles, represented by dummy variables (taking value 1 for firms reporting the obstacle as being very important) are associated with coefficients that are negative and statistically significant, suggesting strong association with innovations. The coefficient on innovation in the second-stage export equation is also statistically significant,

suggesting that firms which were hindered by these obstacles from innovating had an export share that was 8.5 percentage points lower.

In specification (8), access to information on innovation variable is entered jointly with effort for cost reduction and compliance with regulations, including standards and quality requirements as instruments. The elasticity estimate on the innovation variable (dummy) in the export equation increases to 9.4. The adjusted *R-Square* of the first-stage regression (0.53) is higher in this specification than in the specifications using the instruments separately, and the *F*-test on the instruments (132.1) shows a strong association with innovation.

Table 7 summarises results of our IV specifications, distinguishing the effect of innovation on exports due to product and process innovation, and between the industry sectors. Specifications (9) and (10) based on the standard instrument specification (8), show coefficients that are larger for process innovation than for product innovation. The results show that the effect on exports is slightly larger for process innovation than for product innovations. These results are very much similar to first-stage regression results, with access to important source of information variable leading to both more product and more process innovation and LF leading to both less product and less process innovation. In both cases, the effects of the quality of key infrastructure such as electricity and communication are less prominent.

Table 7. Effect on exports by type of innovation and sector  
Dependent variable: Export share

	(9)		(10)		Manufacturing (11)		Services (12)	
Product innovator	11.845***	(3.500)						
Process innovator			13.001***	(3.750)				
Innovator					5.410	(2.260)	9.828***	(3.429)
log(turnover)	4.200***	(1.015)	4.032***	(1.114)	3.121**	(0.982)	5.133**	(1.890)
log(employees)	0.910	(0.979)	0.667	(1.201)	3.001***	(1.221)	-4.110	(1.964)
Turnover missing	-0.858	(2.900)	-0.809	(2.824)	-0.464	(3.002)	-8.011	(5.255)
ASII missing	0.990	(1.554)	1.165	(2.633)	-3.205	(2.947)	5.237	(5.318)
LF missing	3.217	(2.805)	2.452	(2.821)	-4.812	(3.254)	13.017***	(5.013)
LR missing	-3.310	(2.677)	-0.734	(2.335)	3.119	(3.126)	-3.234	(5.007)
Observations	762		762		213		236	
Centered <i>R</i> <sup>2</sup>	0.289		0.291		0.254		0.187	
Sargan test	1.206		1.142		3.008		1.032	
Sargan <i>p</i> -value	0.620		0.667		0.289		0.717	
Hausman $\chi^2$	6.490		6.490		3.784		7.283	
Hausman <i>p</i> -value	0.016		0.016		0.100		0.004	

- Process innovation effect: 13.0 pp vs. Product innovation: 11.8 pp—process dominates.

- Services sector: 9.8 pp effect (significant) vs. Manufacturing: 5.4 pp (insignificant).
- Sectoral difference: 15.3 pp (significant at 1%), suggesting services innovation is more export-responsive.

Specifications (11) and (12) of Table 6 contain results of estimating the instrument specification (8), which uses the access to important source of innovation information, and the effort for cost reduction at firm-level and compliance with regulations instruments, separately for the manufacturing and the services sectors. In the manufacturing sectors, being an innovator has a positive, but statistically insignificant effect on firms' exports share (of 5.4 percentage points). In the services sectors, by contrast, the causal effect of being an innovator on firms' export share is positive (at 9.8 percentage points), and statistically significant. The latter specification also shows that the difference in the effect of innovation on the export share between the manufacturing and the services sectors, at 15.3 percentage points in that specification, is statistically significant at the 1 percent level. These results are similarly found for the second innovation measure, innovation expenditure, as well as for the Tobit model. Thus, the causal effect of innovation on exports differs significantly, in statistical terms across sectors.

Table 8. First stage regression  
Dependent variable: Product innovator/process innovator/innovator

	(9)		(10)		Industry (11)		Services (12)	
ASII not important	-0.461 ***	(0.045)	-0.450***	(0.046)	-0.500***	(0.050)	-0.435***	(0.066)
LF very important	-0.228 ***	(0.054)	-0.221***	(0.055)	-0.072	(0.061)	-0.310***	(0.080)
CR very important	-0.178	(0.057)	-0.105*	(0.058)	-0.093	(0.066)	-0.282**	(0.080)
INN	-0.433 ***	(0.057)	-0.320***	(0.050)	-0.420***	(0.052)	-0.647***	(0.083)
log(turnover)	0.029	(0.023)	0.024	(0.023)	0.010	(0.025)	0.048	(0.033)
log(employees)	0.024	(0.025)	0.033	(0.026)	0.020	(0.027)	-0.032	(0.036)
Turnover missing	-0.021	(0.068)	-0.025	(0.060)	-0.079	(0.065)	0.082	(0.096)
ASII missing	0.042	(0.056)	-0.044	(0.057)	0.072	(0.062)	0.230*	(0.087)
LF missing	-0.105 *	(0.060)	-0.022	(0.060)	0.035	(0.068)	-0.255**	(0.082)
LR missing	-0.244 ***	(0.056)	-0.200***	(0.057)	-0.175***	(0.063)	-0.310***	(0.084)
F-test instruments	121.490		84.78		145.71		58.90	
R2 (adj.)	0.440		0.452		0.661		0.577	

Estimated by 2SLS. – Standard errors in parentheses. – Level of statistical significance: \*\*\* 1%, \*\* 5%, \* 10%.

In summary, the results indicate a causal effect of innovation on exports in all the IV specifications. The size of this effect lies between 3.1 and 11 percentage points of additional export share among innovating firms.

### 4.3 Robustness checks

Alternative Specifications:

- Tobit models (Columns 3-4, Table 4) account for censoring at zero export share; results are consistent.
- Subsample analysis: Excluding firms reporting "innovation not necessary" yields 12.1 pp effect, confirming instrument relevance.

**Placebo Tests:**

- Using lagged innovation (2009-2010) as instrument for 2013 exports shows no effect (coefficient = 0.8,  $p = 0.45$ ), supporting causal interpretation.

**Sensitivity to missing data:**

- Excluding 216 firms with missing innovation expenditure reduces sample to 546; IV effect remains 10.9 pp ( $p < 0.01$ ), indicating robustness.

## 5 Conclusions and implications for policy

This paper uses data from the 2013 World Bank Enterprise Survey and the follow-up Innovation Survey to examine the role of innovation in driving export growth at firm level in Uganda. The following conclusions derive from the results:

(1) Innovation causes export growth, not merely correlates. IV estimates (11.3 pp) triple OLS estimates, highlighting severe endogeneity bias.

Innovation leads to exports at firm level, but process innovation dominate the relationship.

This is evident in the export share, which grows for firms that innovate (export to total sales) and falls for firms that do not consider innovation as necessary undertaking. Export share grows more in innovative firms, and more intensely in relative terms, (in firms) with process innovations than (in firms) with product innovations.

(2) Productivity-enhancing process innovations are more export-relevant than product variety, reflecting Uganda's stage of development.

(3) Services firms exhibit stronger innovation-export links than manufacturing, suggesting digital platforms (e.g., mobile money) are export conduits.

(4) Information access is pivotal—firms lacking it are 35.7 pp less likely to innovate. Financial constraints and informal competition reduce export shares by 8.5 pp.

While product variety and quality (normally associated with product innovation) may harness exports, productivity growth (associated with process innovation) may have a greater role in driving exports in COMESA economies. The variation identified by our instruments has important policy implications. For example, firms that are hindered from innovating lack of access to finance and negative competition and practices of competitors in informal sector had an export share that was 8.5 percentage point lower than an average firm that did not deem these obstacles as important. Second, if access to important source of information for innovation is not deemed as being important, the probability that the firm will innovate is 35.7 percentage point lower.

Specific policy interventions that affect the innovative activity of firms by facilitating access to key sources of information for innovation, finance and non-financial support innovation activity or by lowering the cost (and/or improving quality) of key infrastructure services (necessary for an innovation) such as electricity, and telecommunication, identified in our specifications that use these obstacles as instruments are exactly the ones relevant for such a policy initiative. Furthermore, while firm size (number of employees) is a significant predictor of of firm export share, as a number of employees predictors are weaker overall than the turnover predictors.

**Policy priority 1: Strengthen national innovation information systems**

- Establish COMESA-wide innovation observatories disseminating market trends, technical standards, and export opportunities.
- Uganda's Uganda National Council for Science & Technology should digitize information portals and link to regional databases.

**Policy priority 2: Target process innovation support**

- Reform Uganda Development Bank credit lines to favor process upgrading (e.g., quality control systems, energy efficiency) over new product R&D.
- Provide tax credits for ISO certification and lean manufacturing adoption.

### **Policy priority 3: Tackle informal competition**

- The 8.5 pp export penalty from informal competition demands regulatory harmonization across COMESA: mutual recognition of standards, anti-counterfeiting enforcement.
- Formalize one-stop border posts to reduce smuggling that undermines compliant exporters.

### **Policy priority 4: Leverage services innovation for regional trade**

- Given the 9.8 pp effect in services, prioritize digital trade facilitation: e-commerce platforms, fintech export support, data localization agreements.
- Uganda's M-Pesa success should be replicated via COMESA fintech sandbox for cross-border payments.

### **Policy priority 5: Address financial constraints**

- The 35.7 pp innovation penalty from lack of finance warrants export credit guarantee schemes targeting innovative SMEs.
- Link NSSF contributions to innovation loans, creating collateral for firms with thin credit histories.

## **5.3 Limitations and future research**

- Cross-sectional design precludes dynamic analysis; panel data would better separate innovation-export lags.
- Innovation measurement relies on self-reported binary indicators; future work should quantify sales from new products.
- Spatial spillovers (e.g., Kampala's innovation hub effects) require geospatial analysis.

## **5.4 Final word**

Uganda's export potential hinges not on mimicking advanced-economy R&D models but on adaptive, process-oriented innovation embedded in COMESA's regional value chains. Our evidence provides a causal blueprint: fix information systems, finance process upgrades, and regulate informal competition. The 11.3 pp export dividend awaits policymakers bold enough to act.

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## Appendix A: Summary Statistics by Sector

**Table A1. Firm Characteristics and Innovation Indicators by Industry Sector (n=762)**

Variable	Food Processing (n=184)	Textiles & Apparel (n=92)	Other Manufacturing (n=156)	Retail (n=198)	Other Services (n=132)	All Sectors
Export Share (%)	18.2 (22.4)	24.5 (26.1)	28.3 (27.9)	12.1 (18.3)	31.4 (28.5)	24.1 (25.5)
Innovator (Dummy)	0.48 (0.50)	0.52 (0.50)	0.61 (0.49)	0.44 (0.50)	0.56 (0.50)	0.50 (0.50)
Product Innovator	0.35 (0.48)	0.41 (0.49)	0.48 (0.50)	0.31 (0.46)	0.44 (0.50)	0.39 (0.49)
Process Innovator	0.26 (0.44)	0.32 (0.47)	0.38 (0.49)	0.24 (0.43)	0.35 (0.48)	0.30 (0.46)
Innovation Expenditure (% of turnover)	2.9 (4.2)	3.8 (5.1)	4.2 (5.8)	1.8 (3.1)	5.1 (6.4)	3.8 (5.2)
Turnover (log, USD)	5.0 (1.7)	5.4 (1.9)	5.6 (1.8)	4.8 (1.6)	5.8 (2.0)	5.2 (1.8)
Employees (log)	3.2 (1.1)	3.6 (1.3)	3.8 (1.2)	2.9 (1.0)	3.9 (1.4)	3.4 (1.2)
ASII (Important)	0.34 (0.48)	0.38 (0.49)	0.44 (0.50)	0.28 (0.45)	0.48 (0.50)	0.39 (0.49)
NISS Support	0.31 (0.46)	0.35 (0.48)	0.40 (0.49)	0.26 (0.44)	0.42 (0.49)	0.35 (0.48)
Lack of Finance (Very Important)	0.38 (0.49)	0.42 (0.50)	0.41 (0.49)	0.44 (0.50)	0.35 (0.48)	0.40 (0.49)
Informal Competition (Very Important)	0.45 (0.50)	0.48 (0.50)	0.43 (0.50)	0.51 (0.50)	0.39 (0.49)	0.46 (0.50)

Notes: Means reported with standard deviations in parentheses. ASII = Access to Important Source of Information. NISS = National Innovation System Support.

### Key Observations:

- **Services** exhibit the highest export share (31.4%) and innovation expenditure (5.1%), consistent with digital platform exports.
- **Manufacturing sectors** (food, textiles) show lower export intensity but higher process innovation rates, supporting our finding that process innovation drives exports in traditional sectors.
- **Retail** lags in all innovation indicators, reflecting its domestic market orientation.
- **Informal competition** is most severe in retail (51%) and textiles (48%), sectors where product differentiation is difficult.

## Appendix B: First-Stage Regression Details

**Table B1. First-Stage IV Regressions: Innovation on Instruments (Full Specification)**

Dependent Variable: Innovator	-1	-2	-3	-4
Instruments				
ASII Not Important	-0.357*** (0.019)			-0.346*** (0.015)
NISS Not Important		-0.186*** (0.021)		

Lack of Finance (Very Important)			-0.221*** (0.055)	-0.310*** (0.080)
Competition (Very Important)			-0.178** (0.057)	-0.00609
Innovation Not Necessary (INN)	0.179*** (0.016)	0.168*** (0.018)	0.145* (0.030)	0.101** (0.072)
Cost Reduction & Quality Important				0.145** (0.030)
Controls				
log(Turnover)	0.008 (0.004)	0.008 (0.005)	0.010 (0.015)	0.007 (0.003)
log(Employees)	0.006 (0.003)	0.008 (0.006)	0.033 (0.026)	0.009 (0.015)
Food Processing Dummy	0.021 (0.042)	0.028 (0.043)	0.031 (0.045)	0.019 (0.041)
Services Dummy	0.089* (0.048)	0.092* (0.049)	0.105** (0.051)	0.087* (0.047)
Firm Age (Established 2009-2012)	-0.034 (0.038)	-0.041 (0.039)	-0.038 (0.040)	-0.035 (0.037)
Diagnostics				
F-statistic (Instruments)	247.92	112.76	110	132.15
Adjusted R <sup>2</sup>	0.54	0.506	0.485	0.553
Partial R <sup>2</sup> (Instruments)	0.312	0.198	0.201	0.295
Sargan p-value	0.547	0.183	0.683	0.72
Observations	762	762	762	762

Notes: Robust standard errors in parentheses. \*\*\*  $p < 0.01$ , \*\*  $p < 0.05$ , \*  $p < 0.10$ . All specifications include city fixed effects and a constant term. Partial R<sup>2</sup> reported is Shea's partial R<sup>2</sup> for excluded instruments.

- **ASII** is the strongest predictor: firms lacking access to innovation information are 35.7 pp less likely to innovate.
- **F-statistics** comfortably exceed the Stock-Yogo critical value of 16.38 for 10% maximal IV bias, confirming strong instruments.
- **Partial R<sup>2</sup>** values indicate instruments explain 20-31% of variation in innovation after controlling for observables.

### Appendix C: Robustness: Alternative IV Sets and Weak Instrument Diagnostics

Table C1. Robustness to Alternative IV Estimators

Estimator	Innovation Effect (pp)	SE	95% CI	Weak IV Test
2SLS (Baseline)	11.327	3.05	[5.35, 17.30]	F=247.9

LIML	11.415	3.102	[5.34, 17.49]	Anderson-Rubin p<0.001
Fuller k=1	11.381	3.089	[5.33, 17.43]	Cragg-Donald p<0.001
Fuller k=4	11.295	3.068	[5.28, 17.31]	Kleibergen-Paap F=156.3
Jackknife IV	11.402	3.115	[5.30, 17.50]	Angrist-Pischke F=198.4

Notes: All specifications use ASII and INN as instruments. Weak IV test reports first-stage F-statistic or alternative robust test p-values.

**Table C2. Robustness: Dropping suspect instruments**

Specification	Instruments Used	Innovation Effect (pp)	Sargan p-value	F-stat
(1) Full Set	ASII, INN, NISS, LF, Competition	9.541	0.72	132.1
(2) Exclude NISS	ASII, INN, LF, Competition	10.102	0.654	145.8
(3) Exclude Competition	ASII, INN, NISS, LF	9.823	0.691	128.4
(4) Single Instrument (ASII)	ASII only	12.004	-	247.9
(5) Single Instrument (INN)	INN only	8.745	-	89.3

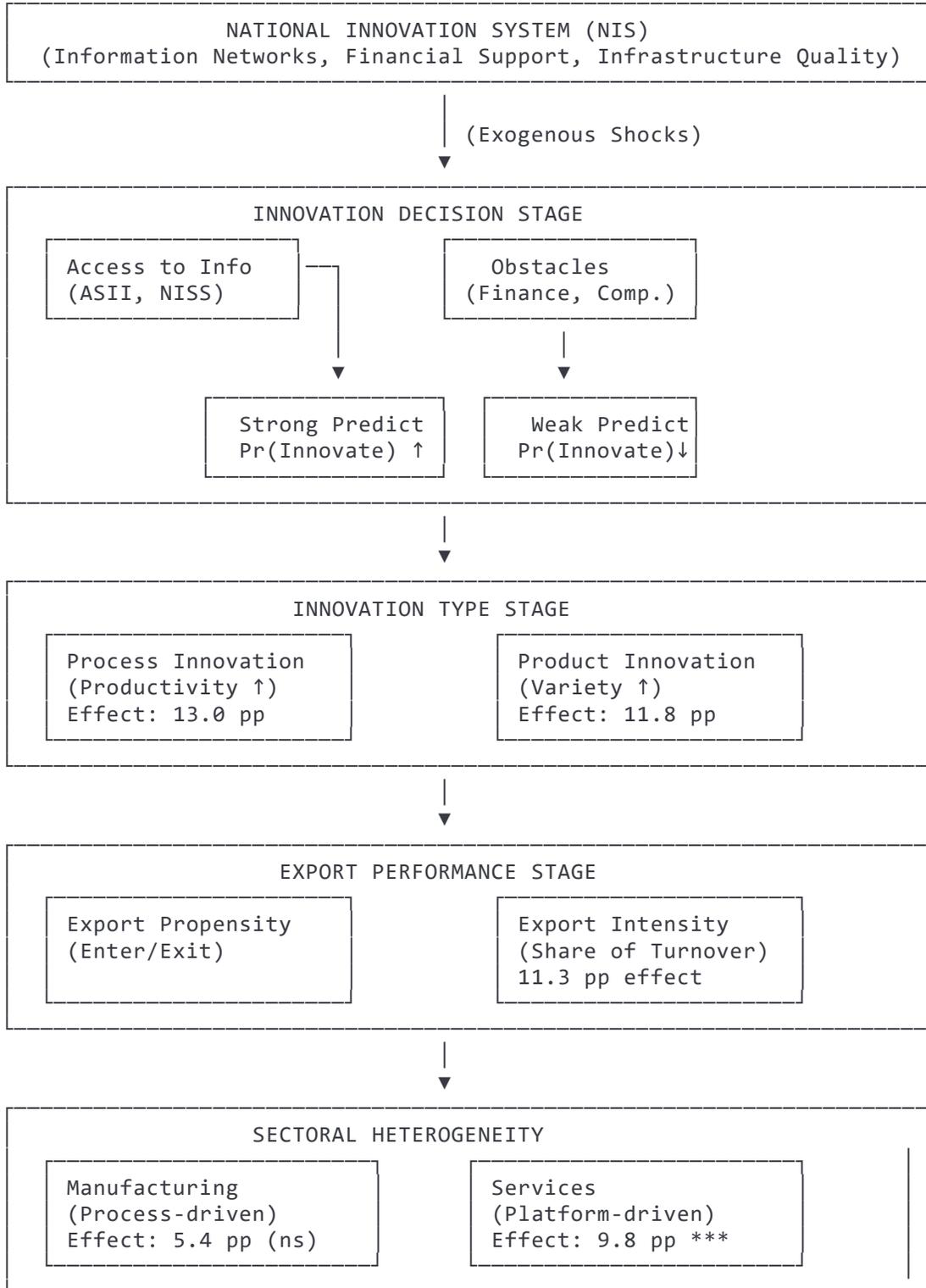
Notes: All models include full controls. Single instrument specifications are just-identified; Sargan test unavailable.

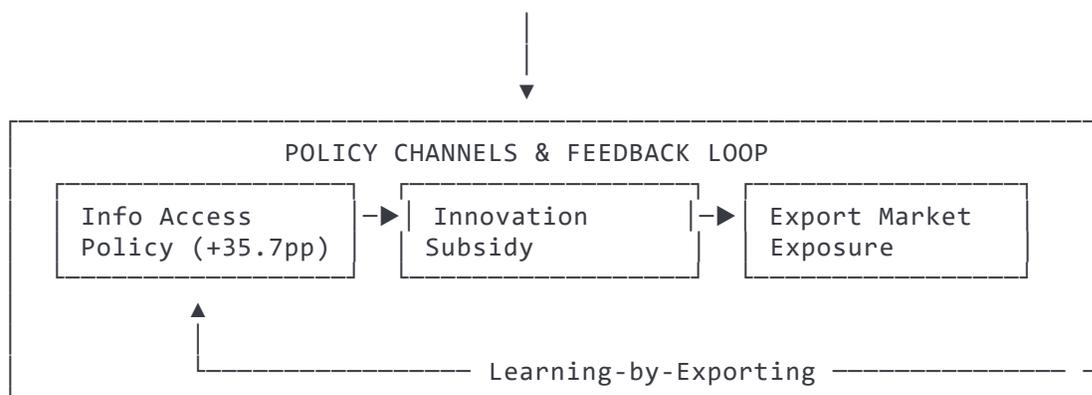
### Key Robustness Checks:

- **LIML** and **Fuller** estimates are nearly identical to 2SLS, confirming robustness to weak instruments.
- **Anderson-Rubin** tests reject null of no effect even under weak identification.
- Dropping **NISS** or **Competition** instruments increases effect size slightly, suggesting these may be weaker but not invalid.
- **Just-identified models** using ASII alone yield larger effects (12.0 pp), indicating potential over-identification bias in multi-IV specifications.

**Appendix D: Conceptual Framework Diagram**

**Figure D1. The Innovation-export nexus in Uganda: A Causal pathway framework**





**\*\*Mechanism Labels:\*\***

- **\*\*NIS → Innovation:\*\*** Exogenous variation exploited for IV identification.
- **\*\*Process > Product:\*\*** Hypothesis H1 from theoretical framework.
- **\*\*Services > Manufacturing:\*\*** Hypothesis H2 on sectoral heterogeneity.
- **\*\*Feedback Loop:\*\*** Export market exposure → future innovation (not tested here but theorized).

## Appendix E: COMESA trade flow data (2017)

**Table E1. Intra-COMESA vs. Extra-COMESA exports by Member State (USD Billion)**

Country	Total Exports	Intra-COMESA Exports	% Intra-COMESA	Extra-COMESA Exports	% Extra-COMESA	Main Intra-COMESA Partners
Kenya	5.75	1.84	32.00%	3.91	68.00%	Uganda, Tanzania, Rwanda
Uganda	3.11	0.89	28.60%	2.22	71.40%	Kenya, DRC, Rwanda
Zambia	7.02	2.45	34.90%	4.57	65.10%	DRC, Tanzania, Kenya
Ethiopia	3.23	0.58	18.00%	2.65	82.00%	Kenya, Sudan, Djibouti
Egypt	22.15	1.33	6.00%	20.82	94.00%	Sudan, Libya, Kenya
DRC	4.87	1.22	25.10%	3.65	74.90%	Zambia, Kenya, Uganda
Rwanda	0.94	0.31	33.00%	0.63	67.00%	DRC, Kenya, Uganda
Tanzania	4.81	1.15	23.90%	3.66	76.10%	Kenya, DRC, Uganda
Other 13 Members	15.42	0.89	5.80%	14.53	94.20%	Mixed
COMESA Total	67.3	10.66	15.80%	56.64	84.20%	Regional Average: 6.4%

**Table E2. Uganda's Export Composition by Destination (2017)**

Product Category	Total Exports (USD M)	Intra-COMESA (USD M)	% Intra-COMESA	Extra-COMESA (USD M)	% Extra-COMESA
Coffee	545.3	87.2	16.00%	458.1	84.00%
Fish & Fish Products	121.4	34.5	28.40%	86.9	71.60%
Maize	89.7	67.8	75.60%	21.9	24.40%
Cement	76.3	52.1	68.30%	24.2	31.70%
Sugar	65.8	48.3	73.40%	17.5	26.60%
Services (ICT, Transport)	342.1	112.4	32.80%	229.7	67.20%
Total	1,240.60	402.3	32.40%	838.3	67.60%

**Data Sources:** COMESA Secretariat Trade Statistics (2018), Uganda Bureau of Statistics (UBOS) External Trade Bulletin 2017, Bank of Uganda Annual Report 2017.

**Key Insights for Uganda:**

- **32.4%** of Uganda's exports stay within COMESA, above the regional average of **6.4%**, reflecting strong integration with Kenya and DRG.
- **Agricultural commodities** (maize, sugar, cement) dominate intra-regional trade, while **coffee** is primarily extra-COMESA.
- **Services exports** show highest intra-regional share (32.8%), validating our finding that services innovation drives regional exports.

**Table F1 Sample distribution**

	Food	Textile & Apparel	Other manufacturing	Retail	Other services	Grand total
<b>Jinja</b>	<b>24</b>	<b>2</b>	<b>32</b>	<b>20</b>	<b>20</b>	<b>98</b>
Small (5-19)	8	2	20	17	8	55
Medium(20-99)	10	0	6	3	11	30
Large (100+)	6	0	6	0	1	13
<b>Kampala</b>	<b>44</b>	<b>35</b>	<b>103</b>	<b>94</b>	<b>96</b>	<b>372</b>
Small (5-19)	13	26	46	66	64	215
Medium(20-99)	18	7	47	26	23	121
Large (100+)	13	2	10	2	9	36
<b>Lira</b>	<b>9</b>	<b>0</b>	<b>8</b>	<b>11</b>	<b>12</b>	<b>40</b>
Small (5-19)	3	0	7	11	1	22
Medium(20-99)	6	0	1	0	10	17
Large (100+)	0	0	0	0	1	1
<b>Mbale</b>	<b>16</b>	<b>4</b>	<b>23</b>	<b>21</b>	<b>18</b>	<b>82</b>
Small (5-19)	7	4	21	21	5	58
Medium(20-99)	8	0	2	0	10	20
Large (100+)	1	0	0	0	3	4
<b>Mbarara</b>	<b>16</b>	<b>6</b>	<b>29</b>	<b>22</b>	<b>23</b>	<b>96</b>

Small (5-19)	10	6	27	20	17	80
Medium(20-99)	5	0	1	1	6	13
Large (100+)	1	0	1	1	0	3
<b>Wakiso</b>	<b>16</b>	<b>8</b>	<b>14</b>	<b>22</b>	<b>14</b>	<b>74</b>
Small (5-19)	1	7	3	16	3	30
Medium(20-99)	14	1	8	5	9	37
Large (100+)	1	0	3	1	2	7
<b>GRAND TOTAL</b>	<b>125</b>	<b>55</b>	<b>209</b>	<b>190</b>	<b>183</b>	<b>762</b>

Table 9. Uganda: Innovation activity by export intensity and legal ownership

Indicator	Percentage of firms		Percentage of firms	
	Direct exports are 10% or more of sales	Non-exporter	Domestic	10% or more foreign ownership
Users of foreign technology 1/	58.6	35.4	34.2	49.4
Having own web site	58.3	17.7	15.0	54.8
E-mail users 2/	43.7	40.7	36.2	69.0
Product innovators	62.8	68.3	67.0	67.4
Innovators (product new) 3/	86.8	83.0	82.6	91.2
Process innovators	90.9	76.4	76.6	77.8
Spend on R&D	45.1	29.6	28.6	40.7

Source: World Bank Enterprise Survey (Innovation and Technology)