

Patterns of Regional and Global Value Chain Participation in the EAC

Sebastian Krantz*

July 29, 2021

Abstract

Using global Multi-Region Input-Output (MRIO) data from 2005-2015, this paper empirically investigates the extent and patterns by which EAC countries have integrated into international production and Global Value Chains (GVCs), and the share of this integration accounted for by Regional Value Chains (RVCs). Results imply that the foreign content of exports (VS, I2E) and the share of exports being re-exported (E2R) are stably between 10% and 20% in most EAC countries. Trade in intermediates with the rest of the world remains 10-12 times greater in value added (VA) terms than trade in intermediates inside the EAC. A significant development in the 2005-2015 period is only visible in Rwanda and Tanzania which have increased their VS (mostly from global suppliers), and in Kenya which has become an important supplier of inputs to the EAC (higher E2R in EAC partners). In addition, a downstream shift is evident across EAC countries and sectors, by which more VA (both domestic and foreign) is used for the production of final goods, while maintaining high levels of exports in primary agriculture and mining. Regressions predicting VA at the sector-level suggest that higher I2E and E2R shares increase GDP with an average elasticity of ≥ 0.25 in the course of 2 years. Estimates for manufacturing sectors were slightly higher at elasticities ≥ 0.3 in response to E2R shifts. These results suggest that policy measures to increase EAC members integration into GVCs and RVCs, and to reverse the downstream trend, are likely to benefit EAC economic growth in the medium run.

Disclaimer

This research was produced out of academic motivation and interest and not commissioned by the Ministry of Finance, Planning and Economic Development or any other EAC policy stakeholder. This manuscript may be privately distributed, but, by virtue of the data used being free for academic use only, not published outside of an academic context. The author considers this manuscript an academic working paper, and intends for its academic publication during subsequent PhD studies. The paper was produced using openly accessible data and open source software packages written for R. Replication materials are available on request to the author at *sebastian.krantz@graduateinstitute.ch*, and will be made openly accessible upon publication.

*ODI Fellow in the Macroeconomic Policy Department, Ministry of Finance, Planning and Economic Development of Uganda, 2020-2021. Previous academic affiliation: Graduate Institute of International and Development Studies (IHEID), Geneva. Academic affiliations from October 2021: Kiel Institute for the Global Economy (IFW) and Christian Albrechts University (CAU), Kiel. Permanent contact: *sebastian.krantz@graduateinstitute.ch*.

Table of Contents

1	Introduction	4
2	Data	5
3	Gross Flows	9
3.1	Intermediate Inputs	10
3.2	Exports	14
3.3	Decompositions	15
4	Value Added Flows	18
4.1	The Leontief Decomposition of Gross Trade Flows	18
4.2	Value Added in Final Production	19
4.3	Intermediate Inputs in Value Added Terms	24
4.4	Value Added Exports	27
4.5	Regional Integration in Value Added Trade	30
4.6	Koopman Wang Wei Decomposition of Gross Exports	33
4.6.1	Aggregate KWW Decomposition	34
4.6.2	Upstreamness and Downstreamness in GVC Participation	34
4.6.3	KWW Decomposition of EAC Exports to the EAC	36
4.6.4	Sectoral-Level Decomposition	37
4.7	New Revealed Comparative Advantage	38
4.7.1	NRCA Relative to the EAC	40
4.7.2	NRCA in Inner-EAC Trade	41
5	GVCs and Industrial Development	42
5.1	GVC Integration and Growth	43
5.1.1	Manufacturing Estimates	49
5.1.2	Comparison of Estimates	49
6	Summary	51
6.1	Gross Flows	51
6.2	Value Added Flows	52
6.3	GVCs and Industrial Development	54
7	Conclusion	56

List of Tables

1	COUNTRIES AND REGIONS	6
2	SECTORS	7
3	LARGEST INTERMEDIATES FLOWS BETWEEN THE EAC AND THE WORLD	11
4	LARGEST INTER-COUNTRY INTERMEDIATE FLOWS WITHIN THE EAC	13
5	LARGEST INTERMEDIATE FLOWS BETWEEN THE EAC AND THE WORLD IN VA TERMS	25
6	LARGEST INTER-COUNTRY INTERMEDIATE FLOWS WITHIN THE EAC IN VA TERMS	26
7	EXCLUDED SECTORS	44
8	SUMMARY STATISTICS OF VARIABLES	44
9	VALUE ADDED REGRESSIONS	47
10	VALUE ADDED REGRESSIONS: ROBUST MM ESTIMATES	48
11	VALUE ADDED REGRESSIONS: MANUFACTURING	49
12	VALUE ADDED REGRESSIONS: MANUFACTURING: ROBUST MM ESTIMATES	50

List of Figures

1	GLOBAL GDP BY REGION	8
2	GLOBAL GDP BY SECTOR	8
3	EAC GDP BY SECTOR	9
4	AGGREGATED MRIO TABLE: EAC AND WORLD REGIONS	10
5	DISAGGREGATED MRIO TABLE: EAC	12
6	GROSS FLOWS RATIOS: ROW/EAC INFLOWS AND OUTFLOWS	13
7	EAC GROSS EXPORTS	14
8	PERCENTAGE OF GROSS EXPORTS GOING TO EAC MEMBERS	15
9	DECOMPOSITION OF OUTPUT AND EXPORTS	16
10	DECOMPOSITION OF SECTORAL OUTPUT AND EXPORTS	17
11	AGGREGATED VALUE ADDED SHARE MATRIX (VB) 2015	20
12	FOREIGN VALUE ADDED SHARES IN EAC PRODUCTION (VS)	21
13	CHANGE IN FOREIGN VALUE ADDED SHARES IN EAC PRODUCTION	22
14	DISAGGREGATED VALUE ADDED SHARE TABLES: EAC IN 2015	23
15	AGGREGATED MRIO TABLE IN VA TERMS: EAC AND WORLD REGIONS	24
16	EAC TRADE BALANCE IN INTERMEDIATE GOODS IN GROSS AND VA TERMS . .	25
17	VA FLOWS RATIOS: ROW/EAC INFLOWS AND OUTFLOWS	26
18	EAC DOMESTIC VALUE ADDED IN GLOBAL EXPORTS	27
19	GVC INTEGRATION OF EAC MEMBERS: AGGREGATE	28
20	GVC INTEGRATION OF EAC MEMBERS: SECTOR LEVEL: 2015	29
21	GVC INTEGRATION OF EAC MEMBERS: ANNUAL GROWTH 2005-2015	30
22	EAC VA SHARES IN MEMBERS VS, VS1, IMPORTS AND FINAL IMPORTS	32
23	KWW DECOMPOSITION OF GROSS EXPORTS	33
24	KWW DECOMPOSITION OF GROSS EXPORTS	34
25	UPSTREAMNESS AND DOWNSTREAMNESS RATIOS	35
26	UPSTREAMNESS AND DOWNSTREAMNESS RATIOS, DIFFERENCE 2005-2015	36
27	KWW DECOMPOSITION OF GROSS EXPORTS TO THE EAC	37
28	KWW DECOMPOSITION OF SECTOR-LEVEL GROSS EXPORTS IN 2015	37
29	UPSTREAMNESS AND DOWNSTREAMNESS BY SECTOR, 2005 AND 2015	38
30	NEW REVEALED COMPARATIVE ADVANTAGE IN 2015	39
31	NRCA ANNUALIZED 2005-2015 GROWTH RATE	40
32	NRCA RELATIVE TO EAC	41
33	NRCA FOR INNER-EAC TRADE	42
34	TIME SERIES OF VARIABLES	45
35	HISTOGRAMS OF VARIABLES	45

1 Introduction

Global Value Chains (GVCs) have become a central topic in trade and development policy. GVCs refer to the quickly expanding internationalization of production networks. While some work has been done on regional value chains (RVCs) in East Africa within specific industries, such as Maize value chains studied by [Daly et al. \(2016\)](#), there has not been a detailed exposition of the GVC participation of East African Countries using comprehensive data sources such as MRIO tables.

One of the few comprehensive analyses of GVCs in Africa is provided by [Foster-McGregor et al. \(2015\)](#) using the EORA 25 sector database over the periods from 2000-2011. They find that Africa as a region is more involved in GVCs than many other developing regions but much of the GVC involvement of Africa is in upstream production, and involves in particular, the supply of primary goods into production of final goods in other regions and countries. Downstream involvement in GVCs is relatively small, and has shown little sign of improving in the last 15 years. Furthermore they find that there is a great deal of heterogeneity in GVC involvement across African countries, with a number of relatively successful countries that are heavily involved in GVCs and with a relatively large share of their involvement being in downstream GVCs.

At a sectoral level, [Foster-McGregor et al. \(2015\)](#) state that manufacturing and high-tech sectors are typically not major contributors to GVC participation in African countries. While manufacturing tends to play a larger role in downstream involvement in GVCs, the agricultural sector still accounts for the largest part of downstream GVC involvement across African countries.

[Foster-McGregor et al. \(2015\)](#) also note that Intra-African GVCs are not particularly important for most African countries, with a number of exceptions in southern Africa. The EU tends to be the biggest GVC partner for Africa, with some evidence to suggest that the contributions of East and South-East Asia, and transition countries are increasing.

The authors also compute various indicators on social upgrading indicating that a minority of African countries have been able to upgrade (prominent examples including Egypt, Nigeria and Tunisia). For most African countries the extent of upgrading is lower than that for the average developing country. Finally, the authors note that due to the overall low volume of exports in some countries, the importance of GVCs may be overstated.

A broader research perspective for developing countries is provided in [Kummritz & Quast \(2016\)](#) and [Kummritz \(2016\)](#). [Kummritz & Quast \(2016\)](#) examine patterns of GVC integration in low-and middle income countries using the OECD TiVA database covering 61 countries and 34 industries for the years 1995, 2000, 2005, and 2008 to 2011. They find that, with exception of the agricultural sector, developing countries are typically located more downstream in the value chain, and export more final goods than high-income countries. They take this as evidence of high-income economies using GVCs to outsource low value added downstream production stages and eventually reimport the final goods. Looking over time they find evidence suggesting that many developing economies have succeeded in moving up the value chain and that the general trend points to a more even distribution of value added across the different countries. Examining different regions, they find that South-East Asia has the highest levels of GVC integration while Latin America and the Caribbean is more heterogeneous with Chile and Costa Rica performing very well. In Africa, Tunisia has developed backward linkages into GVCs, especially with the EU.

Their overall findings suggest that low- and middle-income countries have become an integral part of GVCs, where the foreign content of global value-added exports attributable to these countries has risen from 9% in 1995 to 24% in 2011, and the share in re-exported exports has increased from 9% to 23%. Low- and middle-income countries are increasingly becoming the drivers of GVC expansion and are proceeding up the value chain to more upstream tasks. The authors stipulate that moving upstream should bring greater gains for developing country industrialization.

A more broadly developing country focussed review is also provided by [Kowalski et al. \(2015\)](#).

This paper uses the EORA Global MRIO tables [Lenzen et al. \(2012, 2013\)](#) to analyze patterns

of production in the EAC and compute standard GVC indicators for the years 2005-2015. The goal of this research is to map the structure of regional and international production and exports in the EAC, and to produce some first evidence of the potential benefits of GVC integration for East Africa at the aggregate and sector level.

The analysis will follow for the most part the seminal works of [Hummels et al. \(2001\)](#), [Koopman et al. \(2014\)](#), [Wang et al. \(2013\)](#), as well as [Kummritz \(2016\)](#) and [Kummritz & Quast \(2016\)](#).

2 Data

Most GVC analysis uses Inter-Country Input-Output tables (ICIOs), such as those published by the OECD and WTO (TiVA) or the World Input Output Database ([Timmer et al., 2012](#)). These tables state supply and demand relationships in gross terms between industries within and across countries ([Kummritz & Quast, 2014](#)). The former two databases are however limited to high-income or larger developing countries, with limited or no coverage of Sub-Saharan Africa.

This research therefore uses the EORA Global MRIO tables ([Lenzen et al., 2012, 2013](#)), which have an extensive coverage of 189 countries but rely on more sophisticated supercomputing methods to harmonize data across countries and are therefore considered less reliable than the OECD or WIOD tables.

The EORA database comes in a Full version with heterogenous sector disaggregations as provided by country SUT tables, and an aggregated 26 sector version that is harmonized across countries. This research considers the EORA 26 database, of which data until 2015 is available. Since GVCs are a recent phenomenon, particularly in Africa, and the EAC customs union only became operational in 2005, with Rwanda and Burundi becoming full EAC members in 2007, this research considers the sequence of EORA 26 tables from 2005-2015.

To increase the interpretation of results while preserving some level of detail about the non-EAC world, as well as reduce the strain on computational resources required to obtain results, the non-EAC countries are aggregated into 11 geographic and trade regions as summarized in [Table 1](#). This reduces the size of the transaction tables from $189 \times 25 = 4915$ rows and columns to $(6 + 11) \times 26 = 442$ rows and columns. The 26 sectors are summarized in [Table 2](#)¹.

¹Sector codes are assigned and used throughout the paper, but are not found in the EORA 26 database.

Table 1: COUNTRIES AND REGIONS

<i>Region</i>	<i>Description</i>	<i>Countries</i>
EAC	East African Community	UGA, TZA, KEN, RWA, BDI, SSD
SSA	Sub-Saharan Africa (Excluding EAC)	AGO, BEN, BFA, BWA, CAF, CIV, CMR, COD, COG, COM, CPV, ERI, ETH, GAB, GHA, GIN, GMB, GNB, GNQ, LBR, LSO, MDG, MLI, MOZ, MRT, MUS, MWI, NAM, NER, NGA, SDN, SEN, SLE, SOM, STP, SWZ, SYC, TCD, TGO, ZAF, ZMB, ZWE
EUU	European Union + GBR	AUT, BEL, BGR, CYP, CZE, DEU, DNK, ESP, EST, FIN, FRA, GBR, GRC, HRV, HUN, IRL, ITA, LTU, LUX, LVA, NLD, POL, PRT, ROU, SVK, SVN, SWE, MLT
ECA	Europe and Central Asia (Non-EU)	ALB, AND, ARM, AZE, BIH, BLR, CHE, CHI, FRO, GEO, GIB, GRL, IMN, ISL, KAZ, KGZ, LIE, MCO, MDA, MKD, MNE, NOR, RUS, SMR, SRB, TJK, TKM, TUR, UKR, UZB, XKX
MEA	Middel East and North Africa	ARE, BHR, DJI, DZA, EGY, IRN, IRQ, ISR, JOR, KWT, LBN, LBY, MAR, OMN, PSE, QAT, SAU, SYR, TUN, YEM
NAC	North America and Canada	BMU, CAN, USA
LAC	Latin America and Carribean	ABW, ARG, ATG, BHS, BLZ, BOL, BRA, BRB, CHL, COL, CRI, CUB, CUW, CYM, DMA, DOM, ECU, GRD, GTM, GUY, HND, HTI, JAM, KNA, LCA, MAF, MEX, NIC, PAN, PER, PRI, PRY, SLV, SUR, SXM, TCA, TTO, URY, VCT, VEN, VGB, VIR
ASE	ASEAN	BRN, IDN, KHM, LAO, MMR, MYS, PHL, SGP, THA, VNM
SAS	South Asia	AFG, BGD, BTN, IND, LKA, MDV, NPL, PAK
CHN	China	CHN, HKG, TWN
ROA	Rest of Asia	ASM, GUM, JPN, KOR, MAC, MNG, MNP, NCL, PRK, PYF, TLS
OCE	Oceania	AUS, FJI, FSM, KIR, MHL, NRU, NZL, PLW, PNG, SLB, TON, TUV, VUT, WSM

Table 2: SECTORS

<i>Sector Code</i>	<i>Description</i>
AGR	Agriculture
FIS	Fishing
MIN	Mining and Quarrying
FBE	Food & Beverages
TEX	Textiles and Wearing Apparel
WAP	Wood and Paper
PCM	Petroleum, Chemical and Non-Metallic Mineral Products
MPR	Metal Products
ELM	Electrical and Machinery
TEQ	Transport Equipment
MAN	Other Manufacturing
REC	Recycling
EGW	Electricity, Gas and Water
CON	Construction
MRE	Maintenance and Repair
WTR	Wholesale Trade
RTR	Retail Trade
AFS	Hotels and Restaurants
TRA	Transport
PTE	Post and Telecommunications
FIB	Financial Intermediation and Business Activities
PAD	Public Administration
EHO	Education, Health and Other Services
PHH	Private Households
OTH	Others
REI	Re-export & Re-import

The values recorded in EORA are in thousands of current USD at basic prices². To examine the data, I compute global GDP by region and sector as well as EAC GDP by sector. Figure 1 shows Global GDP by region. The impact of the 2009 global financial crisis is clearly visible and GDP also has declined in 2015. According to this data EAC GDP at basic prices has increased both in absolute value from 43.6 billion USD in 2005 to 101.6 billion USD in 2015, and as a share of global GDP from 0.096% in 2005 to 0.137% in 2015.

Figure 2 shows global GDP by sector. In 2015, 25% of global GDP was produced by financial and business services (FIB), followed by the education, health and other services category at 11.7%. Agriculture and Fishing together only accounted for 4.2% of global GDP in 2015.

²The basic price is the amount receivable by the producer from the purchaser for a unit of a good or service produced as output minus any tax payable, and plus any subsidy receivable, by the producer as a consequence of its production or sale. It excludes any transport charges invoiced separately by the producer

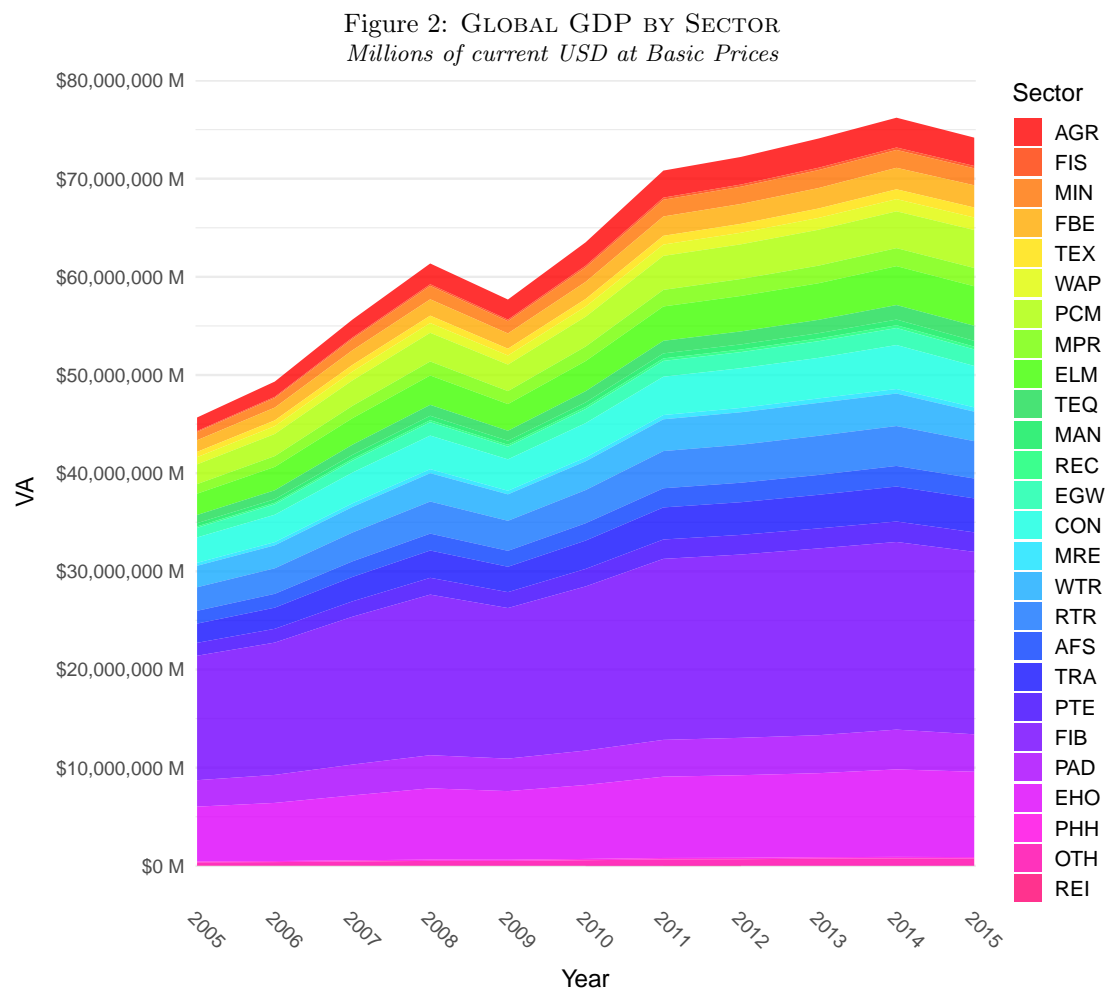
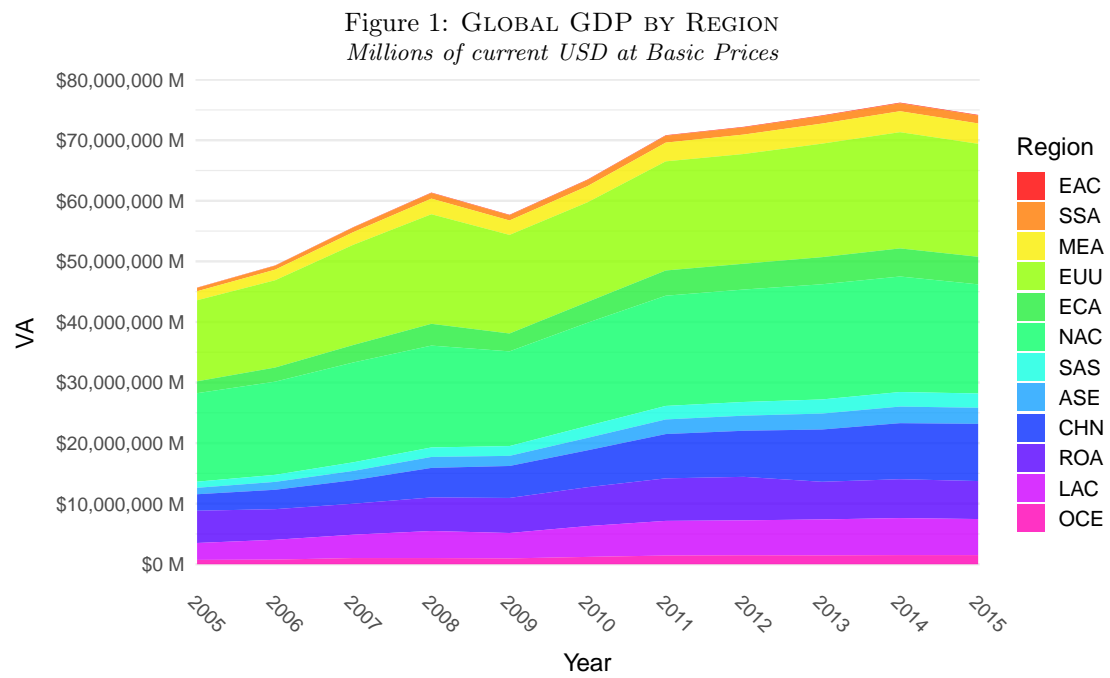
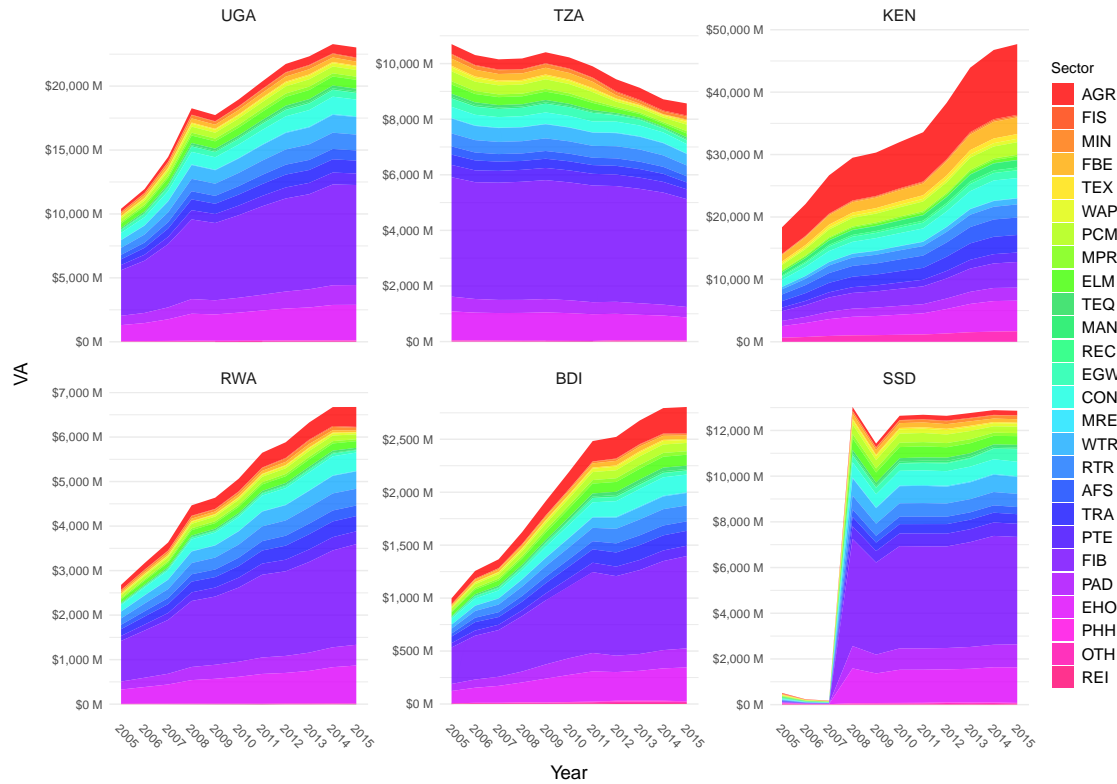


Figure 3 shows EAC GDP by sector. Here discrepancies between this harmonized data and the

real world are very visible. By 2015, agricultural value added in Uganda was still around 30% of GDP, whereas it is below 20% of GDP in the EORA data. The level of GDP at basic prices seems to be broadly in line, as GDP was around 27 billion USD at current prices in 2015, up from 9 Billion in 2005. The growth path however seems to be too strong in the years 2005-2008, and too flat from 2009-2015 compared to the real trajectory. Of the other countries, apart from moderate mismatches in sectoral value added shares for all EAC countries, there seems to be a major problem with the data for Tanzania. Tanzanian GDP was at 18.4 Billion in 2005 and increased to 47.4 Billion in 2015. The EORA data show an initial GDP for Tanzania of 10.5 Billion at basic prices in 2005, which declines over the sample period to 8.5 Billion at basic prices.

Figure 3: EAC GDP BY SECTOR



All of this of course strongly calls into question the reliability of this data to analyze developments in the EAC. The creators of this database write:

The current Eora tables that have been constructed with emphasis on a) representing large data items and b) fulfilling balancing conditions for large countries.

The goal of Eora is to make a consistent global model. When smaller or developing economies have inconsistent or missing data the tables for these countries can become distorted during the process of building a consistent global model.

Thus the analysis and results presented below should be treated with caution, particularly for Tanzania, as the data analyzed was not constructed to accurately reflect macroeconomic aggregates in developing countries. Nevertheless EORA is the only global IO database currently in existence and may be used to get at least a rough idea about production sharing and integration into Global Value Chains in the EAC.

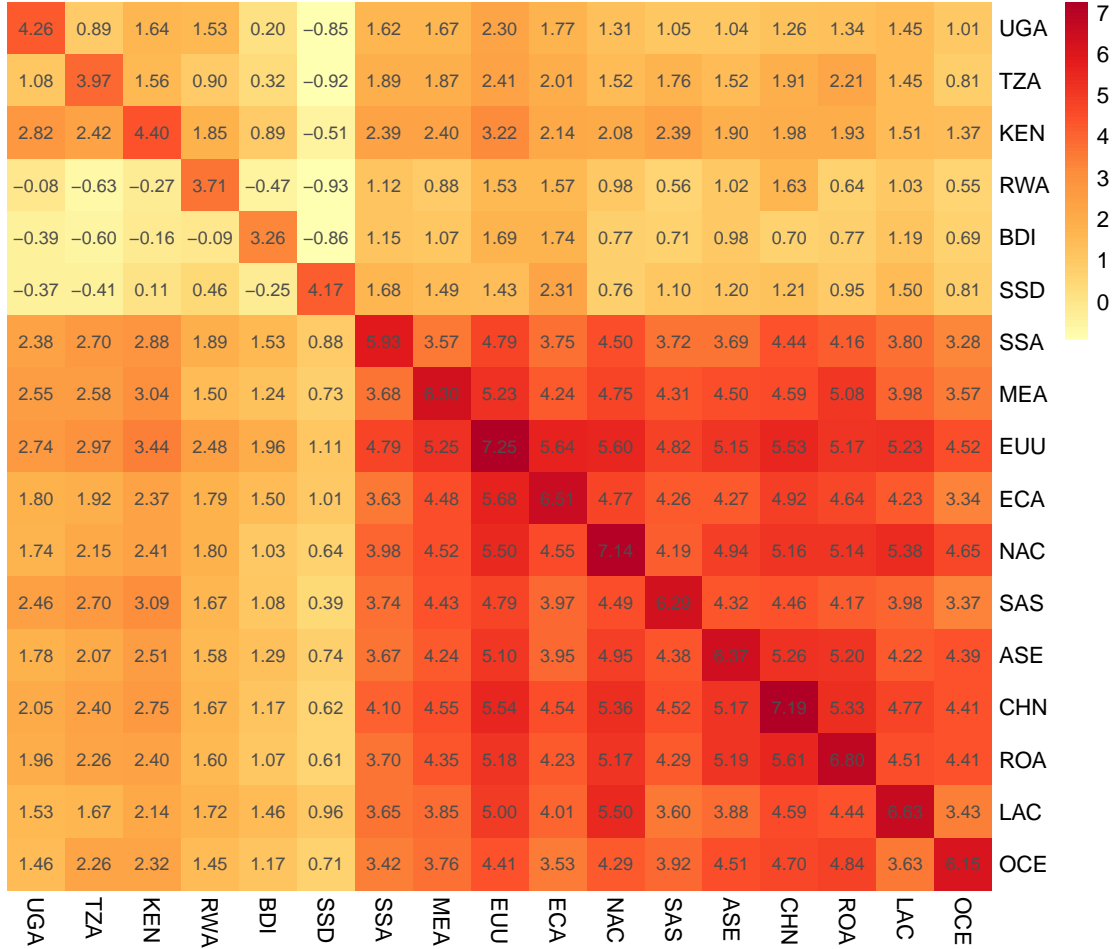
3 Gross Flows

In light of the macroeconomic inconsistencies flagged above and the fact that value added flows are estimated from gross flows, it is useful to first consider the raw data in more detail before diving into detailed decompositions of trade flows.

3.1 Intermediate Inputs

An aggregated EORA 26 MRIO Table for the year 2015 is shown in Figure 4. The columns of the table constitute production functions showing the intermediate inputs required by each of the column-countries or regions from each of the row-countries or regions to produce their output. Conversely the rows show quantities supplied by each row-country or region to each column-country or region. Flows are reported on a log10 scale due to their vastly different magnitudes. Among the EAC countries, the table shows a significant intermediate input supplier role of Kenya, supplying $10^{2.82} = 661$ million USD to Uganda, $10^{2.42} = 263$ million USD to Tanzania and $10^{1.85} = 71$ million USD to Rwanda. Tanzania and Uganda have less of a supplier role with Tanzania supplying 12 million USD to Uganda, 40 million to Kenya and 8 million to Rwanda, and Uganda supplying 8 million to Tanzania, 44 million to Kenya and 34 million to Rwanda. Rwanda appears to be insignificant in terms of its supplier role, supplying less than 1 million USD in inputs to any of its EAC partners. Burundi and South Sudan appear insignificant both as suppliers and consumers of intermediate inputs. With the rest of the World, Uganda, Tanzania and Kenya each import between 250 and 800 million USD of intermediate imports from the rest of Sub-Saharan Africa, and a similar magnitude from the Middle East, South Asia and China. The largest supplier of inputs to each of the EAC countries appears to be the European Union supplying $10^{2.74} = 550$ million USD to Uganda, $10^{2.97} = 993$ million to Tanzania, $10^{3.44} = 2754$ million to Kenya, $10^{2.48} = 302$ million to Rwanda $10^{1.96} = 91$ million to Burundi and $10^{1.11} = 13$ million to South Sudan.

Figure 4: AGGREGATED MRIO TABLE: EAC AND WORLD REGIONS
Millions of 2015 USD at Basic Prices on a Log10 Scale



Visualizing global flows to the EAC at the sector level is not feasible, but below I examine large sector level flows between the EAC and the rest of the World, and then inter-EAC flows. Table 3 shows the 20 largest flows between EAC sectors and sectors outside the EAC, with and without Kenya. The left column shows that the largest flow of almost 460 million USD is Kenyan

agricultural inputs into EU food processing industries, followed by 272 million agricultural goods exports which are just re-exported by the EU. In places 3 and 4 of the largest flows we have inputs from Middle Eastern and EU transport industries to the Kenyan transport industry. The remaining largest flows in the left panel comprise mostly of EU inputs into Kenyan petro-chemicals, construction and electrical machinery. It is also notable that Kenya provides food processing intermediates worth 137 million USD for EU food processing, in addition to the 460 million raw agricultural input and the 272 million re-exported agricultural goods.

If Kenya is taken out, the largest flows, shown on the right hand side of Table 3 are EU inputs to Tanzanian and Ugandan electrical machinery. Tanzania provides agricultural inputs worth 67 million USD to the food processing industries in the rest of Asia (including Japan and South Korea), and Uganda provides agricultural inputs worth 49 million to EU food-processing industries.

Table 3: LARGEST INTERMEDIATES FLOWS BETWEEN THE EAC AND THE WORLD
Millions of 2015 USD at Basic Prices

#	Flow	Value	Non-Kenya Flow	Value
1	KEN.AGR → EUU.FBE	459.214	EUU.ELM → TZA.ELM	128.665
2	KEN.AGR → EUU.REI	271.547	EUU.ELM → UGA.ELM	86.675
3	MEA.TRA → KEN.TRA	186.499	SAS.PCM → TZA.PCM	73.558
4	EUU.TRA → KEN.TRA	178.775	TZA.AGR → ROA.FBE	66.674
5	EUU.ELM → KEN.CON	165.829	EUU.PCM → TZA.PCM	62.432
6	EUU.PCM → KEN.PCM	142.660	MEA.ELM → UGA.ELM	62.200
7	KEN.FBE → EUU.FBE	137.057	SAS.ELM → TZA.ELM	49.312
8	EUU.ELM → TZA.ELM	128.665	UGA.AGR → EUU.FBE	48.568
9	OCE.AGR → KEN.FBE	128.317	SSA.ELM → TZA.ELM	44.663
10	EUU.PCM → KEN.AGR	118.039	SSA.PCM → TZA.PCM	43.131
11	EUU.PCM → KEN.CON	103.888	ROA.WTR → TZA.WTR	41.891
12	EUU.REI → KEN.CON	95.865	MEA.ELM → TZA.ELM	41.537
13	MEA.PCM → KEN.CON	95.677	TZA.AGR → EUU.FBE	39.506
14	EUU.ELM → KEN.ELM	93.319	SAS.ELM → UGA.ELM	37.466
15	SAS.PCM → KEN.PCM	90.327	EUU.ELM → TZA.TEQ	35.433
16	EUU.FBE → KEN.FBE	88.536	EUU.ELM → RWA.ELM	33.555
17	KEN.FBE → EUU.REI	88.051	CHN.ELM → TZA.ELM	31.674
18	EUU.ELM → UGA.ELM	86.675	OCE.ELM → TZA.ELM	31.160
19	SAS.ELM → KEN.CON	82.360	SAS.PCM → UGA.PCM	30.212
20	EUU.PCM → KEN.FBE	77.832	EUU.PCM → UGA.PCM	29.267

A disaggregated view of the inter-EAC part of Figure 4 is presented in Figure 5. It confirms the supplier role of Kenya, particularly in petrol, chemical and mineral products, and other manufacturing sectors such as metal products, electrical machinery, transport equipment as well as trade, transport, telecommunications, financial and business services. As Figure 5 is difficult to interpret in detail, Table 4 additionally records the 20 largest intra-EAC intermediate input flows, with and without Kenyan participation.

Figure 5: DISAGGREGATED MRIO TABLE: EAC
Millions of 2015 USD at Basic Prices on a Log10 Scale



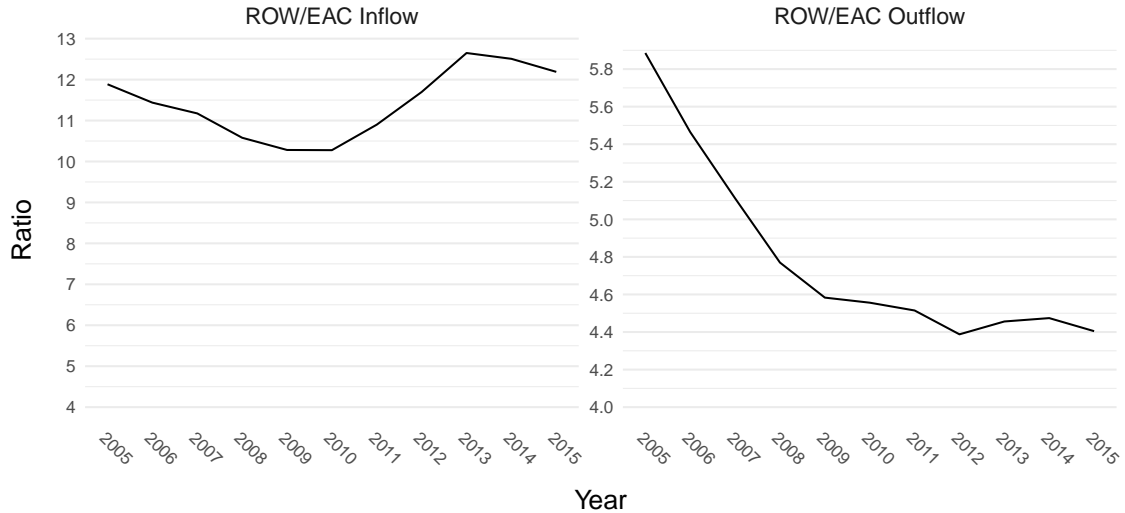
Table 4 shows that the 3 largest inter-country intermediate input flows in the EAC, with values between 37 and 95 million USD, are inputs from Kenyan mining and petrol, chemical and mineral industries to Ugandan and Tanzanian petrol, chemical and mineral industries. The only of the largest 20 inter-EAC flows not originating from Kenya is the flow of Ugandan Agricultural inputs into Kenyan food processing industries, worth 24 million USD. When Kenya is taken out, Uganda becomes the largest supplier of inputs, particularly to Rwandan manufacturing industries. Tanzania supplies mining worth about one million to Ugandan petro-chemicals and food and beverages worth less than 1 million to Ugandan food processing and hotels / restaurants. All other flows in the right-hand side of Table 4 originate from Uganda. Both Rwanda and Tanzania appear to play an insignificant supplier role in the EAC, although it could be dangerous to conclude this about Tanzania given the mismatch of value added shown in Figure 3.

Table 4: LARGEST INTER-COUNTRY INTERMEDIATE FLOWS WITHIN THE EAC
Millions of 2015 USD at Basic Prices

#	Flow	Value	Non-Kenya Flow	Value
1	KEN.MIN → UGA.PCM	95.270	UGA.PCM → RWA.PCM	2.539
2	KEN.PCM → UGA.PCM	63.854	UGA.TRA → RWA.PAD	2.497
3	KEN.PCM → TZA.PCM	37.412	UGA.MPR → RWA.MPR	2.091
4	KEN.WAP → UGA.WAP	29.109	UGA.TRA → RWA.TRA	2.003
5	KEN.ELM → UGA.ELM	25.912	UGA.FBE → RWA.FBE	1.958
6	UGA.AGR → KEN.FBE	24.319	UGA.MPR → RWA.ELM	1.443
7	KEN.TRA → UGA.PAD	23.140	UGA.ELM → RWA.ELM	1.346
8	KEN.PCM → UGA.EHO	20.892	UGA.FBE → RWA.AFS	1.175
9	KEN.TRA → UGA.TRA	20.085	UGA.WTR → RWA.WTR	1.124
10	KEN.MIN → UGA.EGW	18.863	UGA.PCM → TZA.PCM	1.088
11	KEN.MIN → TZA.PCM	18.044	TZA.MIN → UGA.PCM	0.992
12	KEN.WAP → TZA.WAP	15.156	UGA.AGR → RWA.FBE	0.824
13	KEN.FBE → UGA.FBE	14.913	UGA.PCM → RWA.EHO	0.817
14	KEN.WAP → UGA.CON	14.288	UGA.WAP → RWA.WAP	0.813
15	KEN.MPR → UGA.ELM	13.857	TZA.FBE → UGA.FBE	0.742
16	KEN.PCM → TZA.EHO	11.961	UGA.ELM → TZA.ELM	0.631
17	KEN.ELM → UGA.MPR	11.708	UGA.MPR → RWA.CON	0.535
18	KEN.ELM → TZA.ELM	11.688	UGA.MPR → RWA.TEQ	0.479
19	KEN.ELM → UGA.TEQ	11.555	TZA.FBE → UGA.AFS	0.471
20	KEN.PCM → UGA.PAD	11.140	UGA.PCM → RWA.PAD	0.453

To better understand the relative magnitude of IO relationships within the EAC vis-a-vis the rest of the World (ROW), Figure 6 plots the relative magnitude of ROW inflows into EAC production to EAC inputs into EAC production (excluding own inputs), and next to it the ratio of EAC inputs into ROW production divided by EAC inputs into EAC production.

Figure 6: GROSS FLOWS RATIOS: ROW/EAC INFLOWS AND OUTFLOWS



It is evident from Figure 6 that the IO relationships of the EAC with the ROW have developed asymmetrically. ROW continues to supply 12 times more inputs for EAC production than other EAC members, but the ratio of EAC inputs for ROW production to EAC members inputs for EAC production has declined from 5.9 in 2005 to 4.4 in 2015. This implies a relatively greater demand for EAC inputs from EAC members compared to ROW, but an inability of EAC countries to increasingly meet the demands of their production with inputs sourced from EAC members.

3.2 Exports

Gross exports of EAC countries are shown in Figure 7. Here the level of Tanzanian exports is more in line with the level recorded by the World Bank. In terms of composition, it is evident that Uganda focuses on agricultural exports, comprising 38% of exports over the analyzed period, while Rwanda has a disproportionate share in mining exports of about 24%. The other EAC countries have a more balanced export mix, with Tanzania and Kenya also maintaining shares of 24% and 29%, respectively, in agriculture.

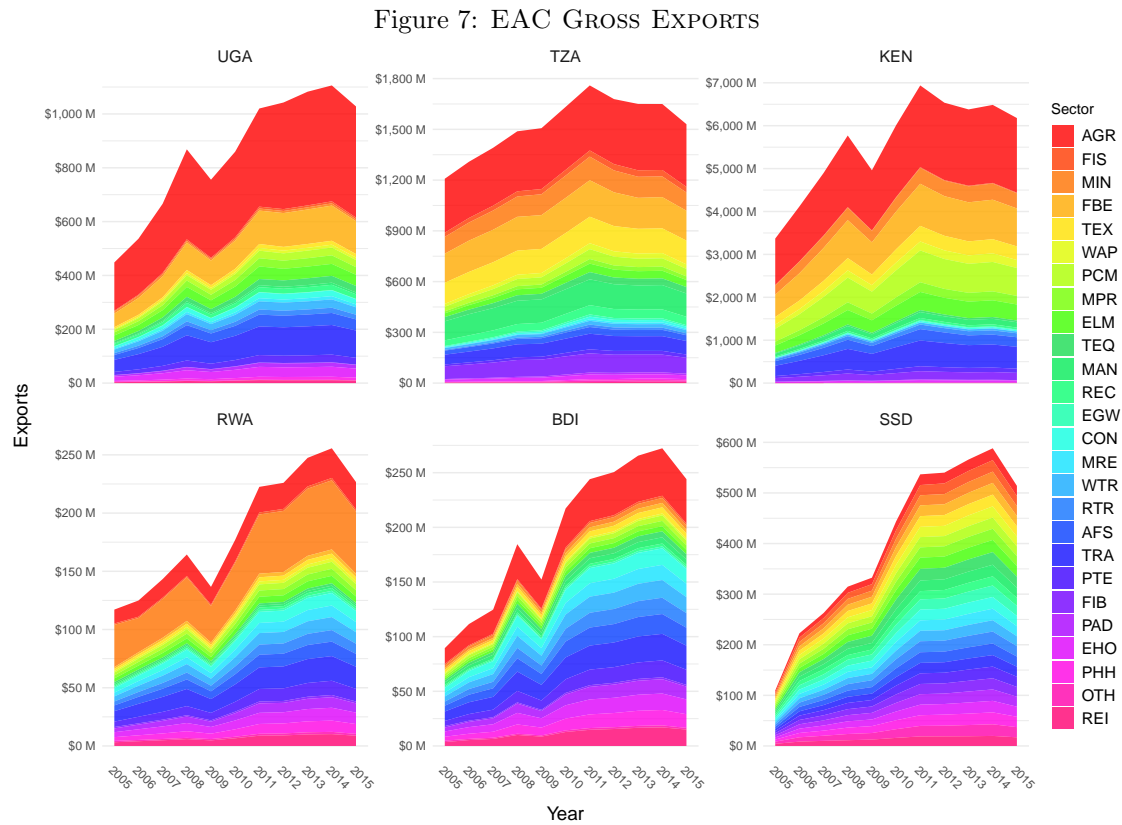
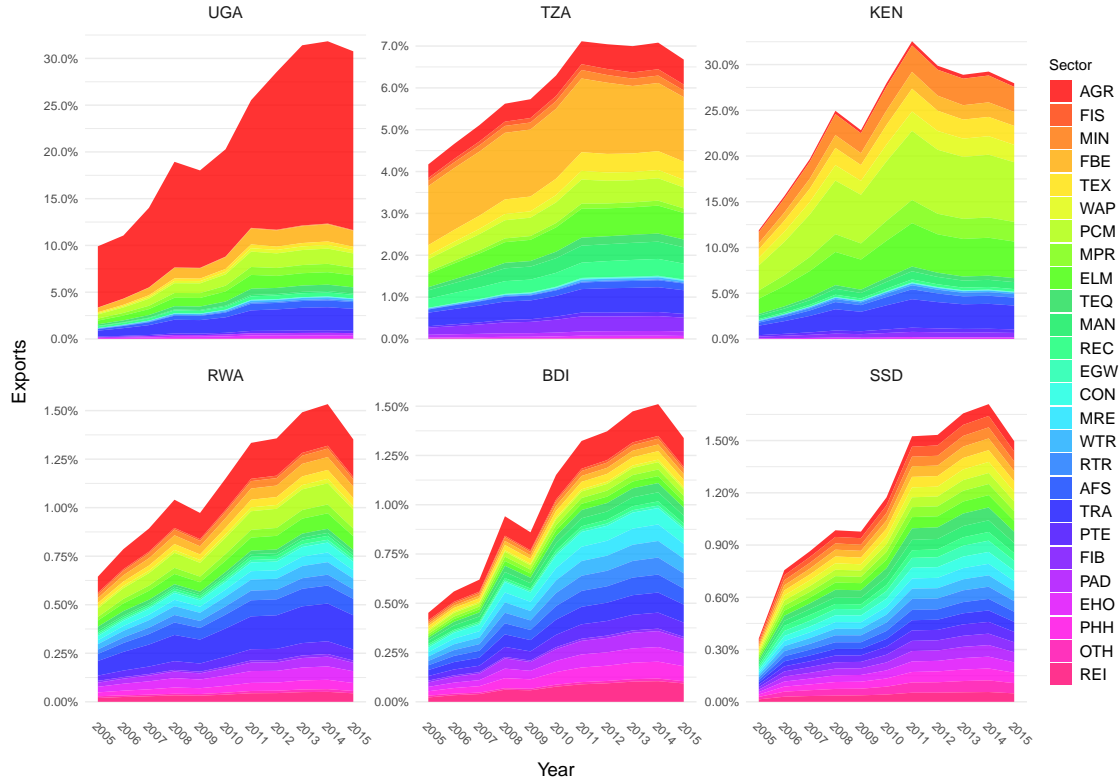


Figure 8 shows the percentage of gross exports going to EAC member countries. It is evident that Uganda and Kenya both have shares of 30% of their exports to the EAC, and that for Uganda the largest part of these exports are agricultural, while for Kenya the largest part is manufacturing, in particular petro-chemicals, metal products and electric machinery. The other EAC members don't export very much to the EAC, in particular Rwanda, Burundi and South Sudan where the data suggests an EAC export share below 2% in 2015.

Figure 8: PERCENTAGE OF GROSS EXPORTS GOING TO EAC MEMBERS

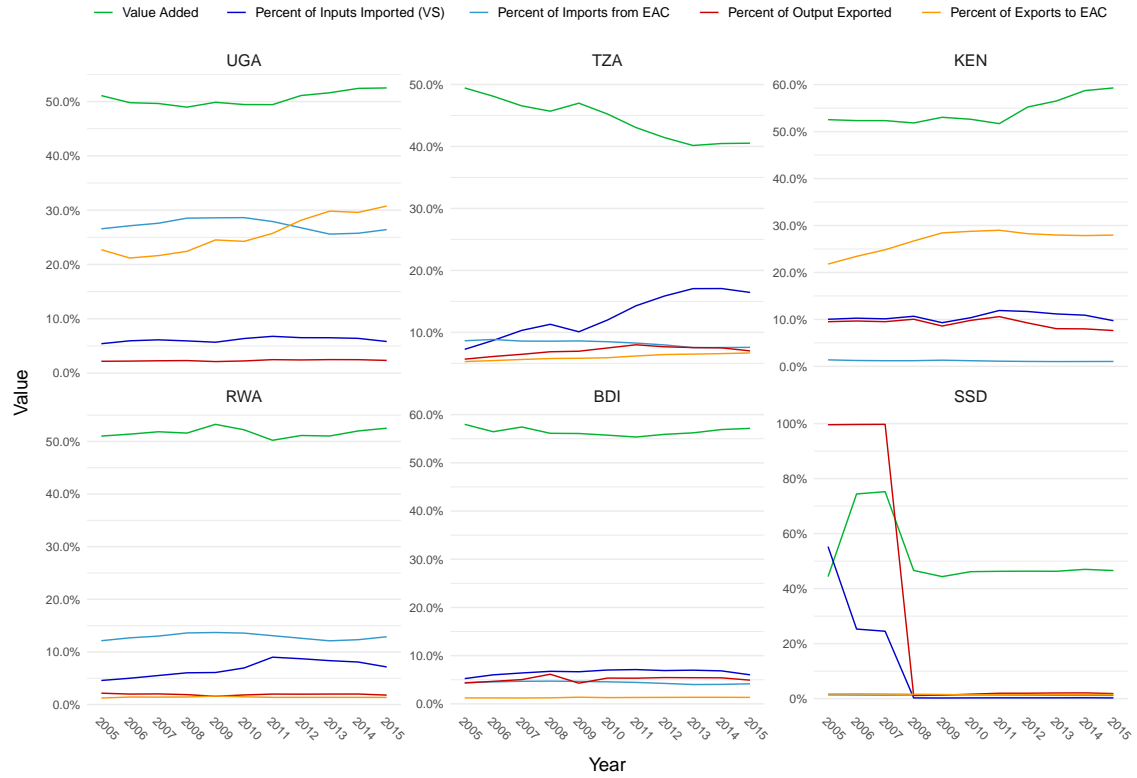


3.3 Decompositions

Figure 9 shows an aggregate decomposition of output and exports into import and export shares, and shares from the EAC. Value Added gives the total share of domestic value added in output (VAS), which appears to be stable around 50-60% for all countries apart from Tanzania where it seems to have dropped down to 40% (which might be due to inconsistencies in domestic data for Tanzania). The remainder of output (1-VAS) is comprised of domestic or imported intermediate goods. The Percent of Inputs Imported gives the share of intermediate inputs that is imported. It is a gross measure of vertical specialization and backward GVC integration, although not the measure proposed by [Hummels et al. \(2001\)](#), which is defined in value added terms. Of those imported inputs, the Percent of Imports from EAC shows the percentage coming from the EAC. It is a measure of backwards regional integration relative to the overall level of a countries international integration. Similarly on the export side Figure 9 reports both the overall percentage of output exported (i.e. not consumed domestically) and the percentage of exports going to the EAC neighbours.

The most curious finding presented by Figure 9 is the remarkable stability of shares, with few exceptions, suggesting only a very moderate increase in regional and global economic supply chains at the aggregate level. The starting levels of the countries are very different, with Uganda maintaining shares of exports and imports around 30% with the EAC, whereas Tanzania, Rwanda and Burundi show much lower levels of integration. In Uganda and Kenya the percent of exports going to the EAC increased slightly over the sample period, while maintaining stable imported inputs and exported output shares. The overall increase in the percent of inputs imported in Tanzania may be due to a decline in domestic intermediates reflecting the decline in GDP, and should thus also be taken with extreme caution.

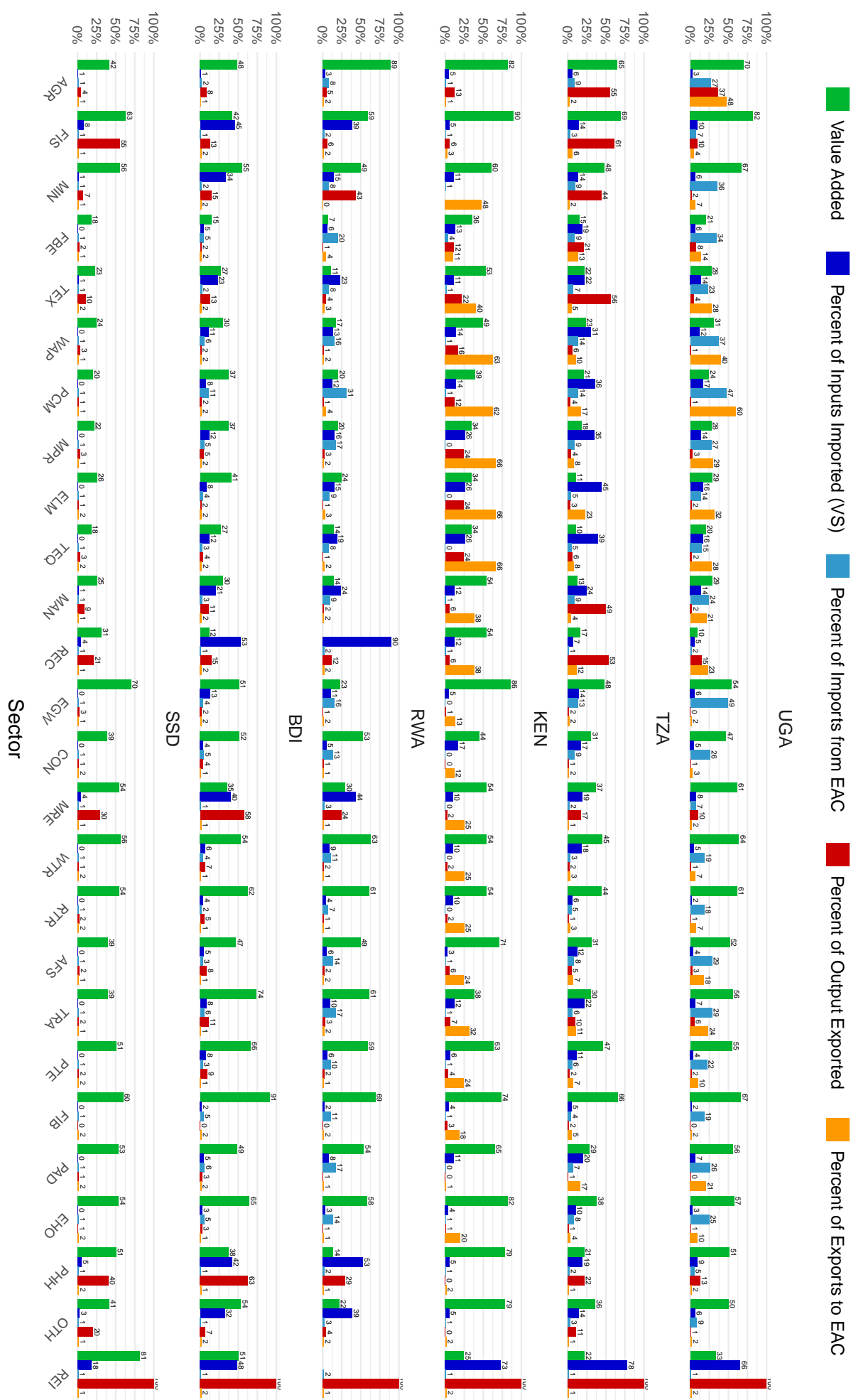
Figure 9: DECOMPOSITION OF OUTPUT AND EXPORTS



An industry level view of these metrics for the year 2015 is provided in Figure 10. It shows some similarities but also some quite strong differences in the structure of production of different EAC countries. For example Agriculture is a high domestic value added and low imported inputs sector in all EAC countries. Uganda and Tanzania export 37 and 55% of their Agricultural produce, while Kenya only appears to export around 13% of its production, and the remaining EAC countries have agricultural export shares below 10%. A major difference between Uganda and Tanzania is that Uganda obtains 27% of its imported inputs in agriculture from the EAC, and sends 48% of its agricultural exports to the EAC, whereas Tanzania only imports 9% of imported intermediates from the EAC and only exports 2% of its agricultural exports to the EAC.

A pattern visible across most manufacturing sectors and supporting services is that Uganda maintains overall lower export and import shares in its production, but in those has significant shares of 20-40% with its EAC partners. Tanzania appears to have high shares of imported intermediates in its core manufacturing sectors, ranging from 30-45% of inputs being imported. Only around 5-15% of these imports are however from the EAC. The sectoral data also show that Tanzania does not export much of its manufacturing output, with the exception of textiles (56%), other manufacturing (49%) and recycling (53%). Thus Tanzania does not appear to engage in significant processing trade but produces manufactured goods with high imported content mostly for domestic consumption. In Kenya, the domestic value added shares in manufacturing are higher and the share of imported intermediates lower than in Tanzania, with around 10-25% of intermediates imported in the core manufacturing sectors. In contrast to Tanzania, Kenya also exports around 15-25% of its output, and most of these exports (around 60%) are to its EAC neighbours. Thus Kenya figures as an important supplier of manufactured goods in the EAC. Rwanda imports around 10-20% of its manufacturing inputs of which about 10-30% are obtained from its EAC neighbours. Rwanda however exports only a negligible fraction of its manufacturing outputs, and of these exports also only a negligible amount goes to the EAC.

Figure 10: DECOMPOSITION OF SECTORAL OUTPUT AND EXPORTS



To summarize, the decomposed gross flows data presented in Figures 9 and 10 suggest that economic integration from the production side has proceeded very gradually, both with the rest of the World and inside the EAC. The various EAC countries take on quite different roles in this process, with Kenya being the major EAC supplier of manufacturing inputs, and Tanzania the country that imports most inputs from abroad to produce for domestic consumption. Uganda shows modest amounts of overall economic integration, but retrieves a significant share of its imported inputs from the EAC, and also exports a significant share of its exports to the EAC - in particular agricultural exports feeding into Kenyan food processing industries. Rwanda and Burundi import around 10-15% of their intermediates, of which about 10-20% come from the EAC. These countries hardly export any intermediate goods. South Sudan does not appear to be economically integrated with EAC production.

4 Value Added Flows

While gross flows provide useful information about direct productive relationships and the amounts of goods traded therein, a problem of gross flows IO tables, is that they do not reveal how much of the value was added in the supplying industry, and how much of the value was added in previous stages of production, performed by other industries or even countries. The Leontief decomposition of gross trade flows solves this problem by reallocating the value of intermediate goods used by industries to the original producers (Kummritz & Quast, 2014).

4.1 The Leontief Decomposition of Gross Trade Flows

Let \mathbf{A} be a row-normalized ICIO table where each element a_{ij} gives the units of sector (row) i 's output required for the production of one unit of sector (column) j 's output, \mathbf{x} the vector of outputs of each country-industry and \mathbf{d} a vector of final demands such that the following productive relationship holds:

$$\mathbf{x} = \mathbf{A}\mathbf{x} + \mathbf{d}. \quad (1)$$

The classical Leontief (1936) insight was that one can solve this equation for \mathbf{x} to get the amount of output each industry should produce given a certain amount of final demand:

$$\mathbf{x} = (\mathbf{I} - \mathbf{A})^{-1}\mathbf{d} = \mathbf{B}\mathbf{d}, \quad (2)$$

where the Leontief Inverse was denoted $\mathbf{B} = (\mathbf{I} - \mathbf{A})^{-1}$. This matrix is also often called the total requirement matrix since it gives the total productive input requirement from each sector to produce one unit of final output. Specifically each element in b_{ij} in \mathbf{B} gives the output required from sector i for the production of one unit of the final good j . Thus the first column of \mathbf{B} gives all the productive input required from all sectors for the production of one unit of the final good in sector 1, and the first row of \mathbf{B} gives all the input required from sector 1 to produce one unit of the final good in each sector. Now the amount of direct value added in each unit of output for each sector is given by:

$$\mathbf{v} = \mathbf{1} - \mathbf{A}'\mathbf{1} \quad (3)$$

where $\mathbf{1} = (1, 1, 1, \dots, 1)'$ is a column-vector of 1's such that the above expression amounts to summing up the entries in each column of \mathbf{A} (representing the intermediate input shares for 1 unit of output) and subtracting them from 1. Let \mathbf{V} be the matrix with \mathbf{v} along the diagonal and 0's in the off-diagonal elements. Multiplying Eq. 2 with \mathbf{V} therefore gives the value added in each sector:

$$\mathbf{V}\mathbf{x} = \mathbf{V}(\mathbf{I} - \mathbf{A})^{-1}\mathbf{d} = \mathbf{V}\mathbf{B}\mathbf{d}. \quad (4)$$

The term $\mathbf{V}\mathbf{B} = \mathbf{V}(\mathbf{I} - \mathbf{A})^{-1}$ is known as the matrix of value added multipliers or value added shares, which can be used to obtain the amount of value added generated in each industry ($\mathbf{V}\mathbf{x}$) when producing to satisfy final demand (\mathbf{d}). More generally, the matrix $\mathbf{V}\mathbf{B}$ contains the amount of value added by each sector (row) to the production of one unit of each sector's (column's) output. To see this, note first that it can be proven that the columns of $\mathbf{V}\mathbf{B}$ sum to 1:

$$(\mathbf{VB})'\mathbf{1} = \mathbf{B}'\mathbf{V}'\mathbf{1} = \mathbf{B}'\mathbf{v} = (\mathbf{I} - \mathbf{A}')^{-1}(\mathbf{1} - \mathbf{A}'\mathbf{1}) = (\mathbf{I} - \mathbf{A}')^{-1}(\mathbf{I} - \mathbf{A}')\mathbf{1} = \mathbf{1}. \quad (5)$$

It can also easily be proven that the sum $(\mathbf{Vx})'\mathbf{1}$ is equal to the sum $\mathbf{d}'\mathbf{1}$: Using Eq. (1) yields $\mathbf{d} = \mathbf{x} - \mathbf{Ax}$, and Eq. (3) yields $\mathbf{Vx} = \mathbf{vx} = \mathbf{x} - \mathbf{A}'\mathbf{x}$, so that $(\mathbf{Vx})'\mathbf{1} = \mathbf{x}'\mathbf{1} - \mathbf{x}'\mathbf{A}\mathbf{1}$ and $\mathbf{d}'\mathbf{1} = \mathbf{x}'\mathbf{1} - (\mathbf{Ax})'\mathbf{1}$. Now since \mathbf{A} is row-normalized by output \mathbf{x} , \mathbf{Ax} simply gives the vector of intermediate use of each industries output, which is summed across industries through transposing and multiplication by $\mathbf{1}$. $\mathbf{A}\mathbf{1}$ gives the vector of intermediate shares in each industries output, which is converted to quantity and summed through pre-multiplying by \mathbf{x}' .

Thus Eq. 4 actually gives us a transformation of the finally demanded quantities \mathbf{d} to the value added origins of those same quantities \mathbf{Vx} . As mentioned before, each element of \mathbf{VB} represents the share of value added of the row country-industry to the production of one unit of the column country-industry's final product.

Another way to arrive at this, following Wang et al. (2013), is to consider the value added generated by producing one unit of final output in each sector. The direct value added at the final stage of production in each sector is \mathbf{V} , but this would exclude the value added by suppliers supplying intermediate inputs to this output. The additional value added by other sectors supplying intermediate inputs to this final output is \mathbf{VA} ³, which needs to be added to the final value added \mathbf{V} . Now in order to generate inputs, the supplying sectors also require intermediate inputs from other sectors, of magnitude \mathbf{AA} , generating a value added \mathbf{VAA} . The process continues through many rounds of production needed to produce those final intermediates needed to produce one unit of output, and the value added generated can be expressed as an infinite geometric series:

$$\mathbf{V} + \mathbf{VA} + \mathbf{VAA} + \mathbf{VAAA} + \dots = \mathbf{V} \sum_{i=0}^{\infty} \mathbf{VA}^i = \mathbf{V}(\mathbf{I} - \mathbf{A})^{-1} = \mathbf{VB} \quad \text{as } a_{ij} \leq 1 \quad \forall i, j. \quad (6)$$

Thus \mathbf{VB} is the value added matrix which sums up the total value added generated by each sector both as a final producer and as a supplier of inputs that is needed to produce one unit of final output in each sector.

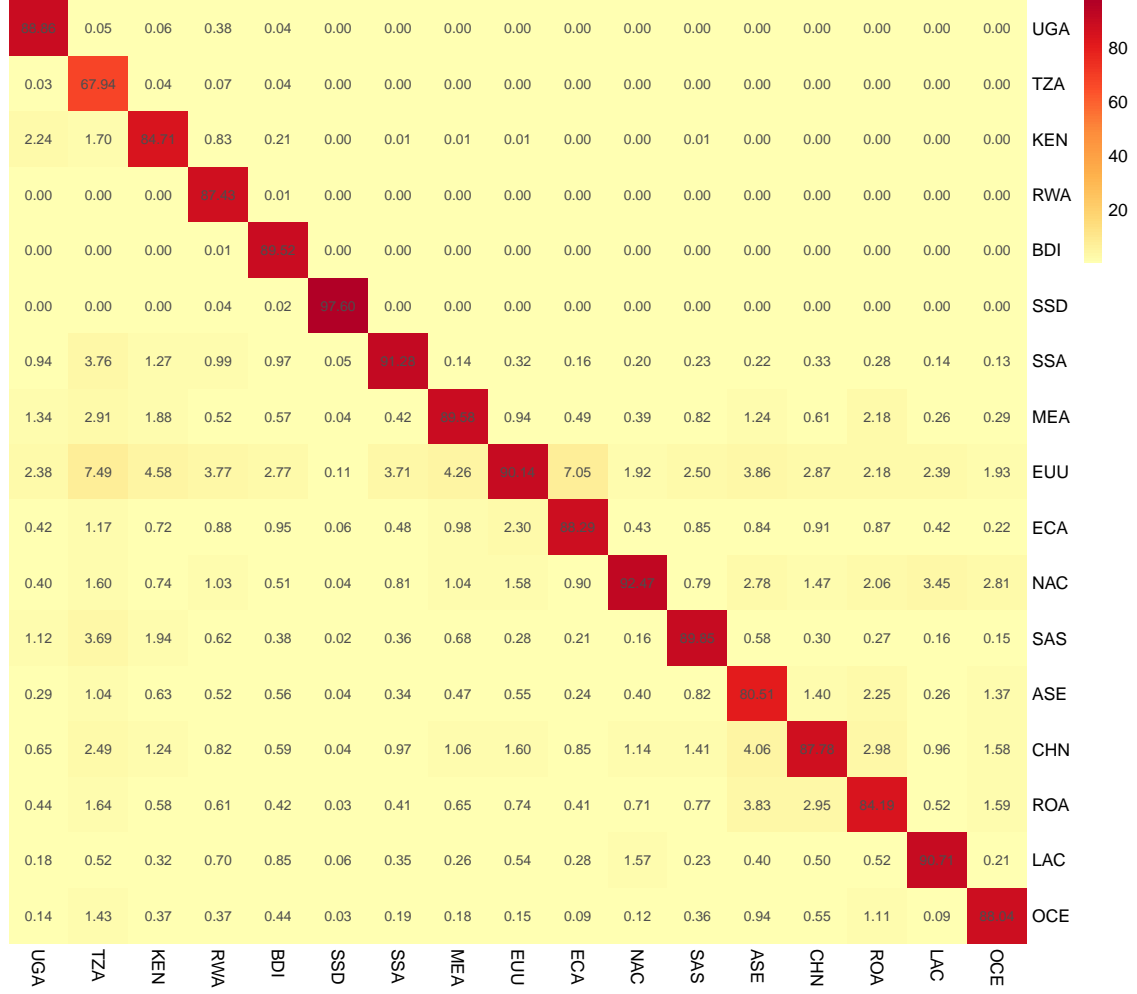
If we assume that the production technology of goods is the same no matter whether they are domestically consumed or exported, then we may also apply the matrix \mathbf{VB} to exports to obtain the value added origins thereof. Before analyzing quantities, let us visualize the matrix \mathbf{VB} at both the aggregate and EAC disaggregated levels.

4.2 Value Added in Final Production

Figure 11 shows the aggregate value added share matrix at the EAC country and world region level. In most EAC countries (Uganda, Kenya, Rwanda and Burundi), approx. 85-90 percent of the value is added by domestic producers. Notable exceptions are Tanzania with 69 percent domestic value added and South Sudan with 98 percent domestic value added. On the supply side, the EU is the largest supplier of intermediate inputs to the production of most EAC countries, followed by South Asia, the Middle east and Kenya, with significant supplier roles also taken by China, North America, and the rest of Sub-Saharan Africa. Within the EAC, Kenya clearly takes on the largest supplier role, adding 2.24 percent of the value in Ugandan production, 1.7 percent in Tanzanias production and 0.83 percent in Rwanda's production. Of the other EAC nations, only Uganda seems to play a non-negligible supplier role for Rwanda, where it supplies 0.38 percent of value added in Rwandan production.

³ \mathbf{A} gives the units of intermediate input required from each sector (row) for the production of one unit of the sector's (column's) output.

Figure 11: AGGREGATED VALUE ADDED SHARE MATRIX (**VB**) 2015
Shares in Percentage Terms, Columns Sum to 100 Percent



The foreign value added share in domestic production and exports is what [Hummels et al. \(2001\)](#) termed 'Vertical Specialization' (VS), and is the most widely used measure of backward GVC integration. Consider the VA share matrix **VB** with elements $vb_{oi,dj}$ where o is the VA origin country and i the VA origin industry (along the rows) and d is the VA using country and j the VA using industry (along the columns). Then the VS ratio for a particular country-industry may be expressed as:

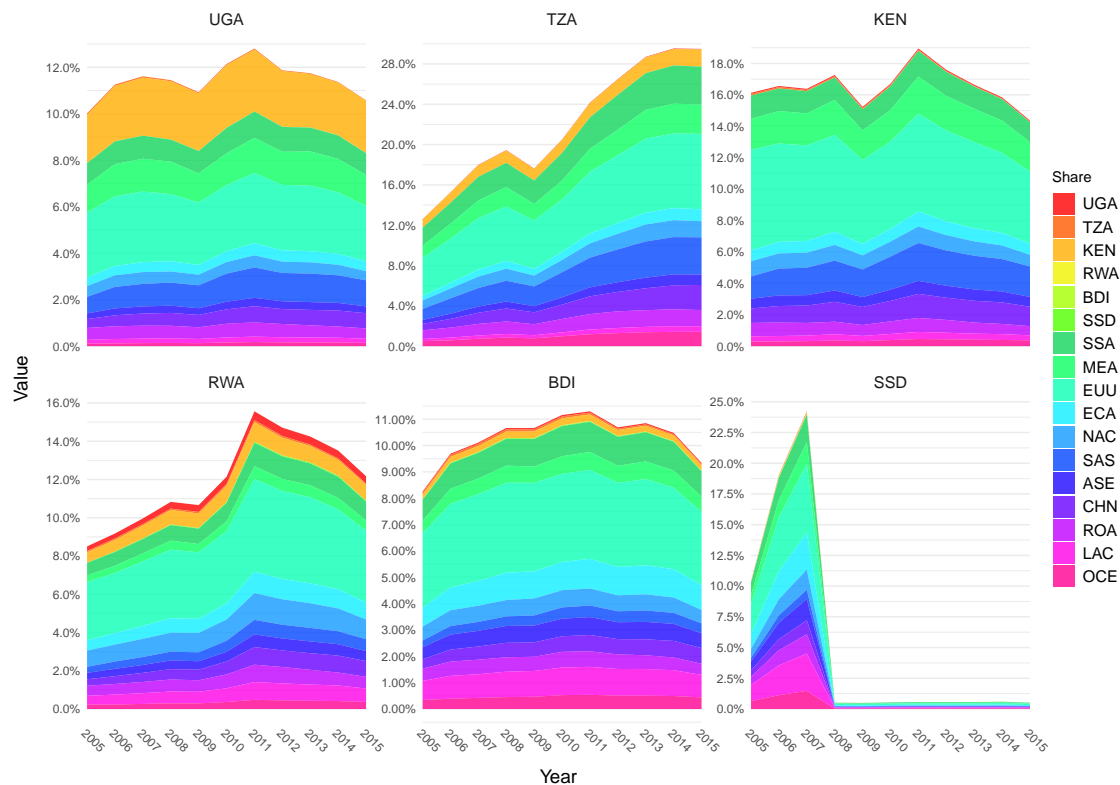
$$VS_{uj} = \sum_{oi, o \neq u} vb_{oi,uj} \quad \forall uj, \quad (7)$$

in other words we are summing the elements of **VB** in each column, excluding any VA shares components from domestic country-industries. Figure 12 gives a breakdown of the VS for EAC countries by supplier country-region over the sample period. When comparing it to the foreign intermediates share in gross flows shown in Figure 9, it is evident that the VS computed from value added data is about 1.5 times greater in nearly all EAC countries, owing to the fact that the foreign content in domestic intermediates which is now taken into account is a lot greater than the domestic content in foreign intermediates which is subtracted in value added flows data.

Figure 12 shows that the share of foreign value added in Ugandan production has been fluctuating between 10 and 12% over the analyzed period. 2% of the value of Ugandan produce comes from Kenya, about 1% from the rest of Sub-Saharan Africa, 1.5% from the MENA region, about 3-4% from the EU and 1-1.5% from South Asia. The other regions make up the remaining 2%. Tanzania and Kenya have similar relative VA contributions of SSA, MENA, EU and SAS regions

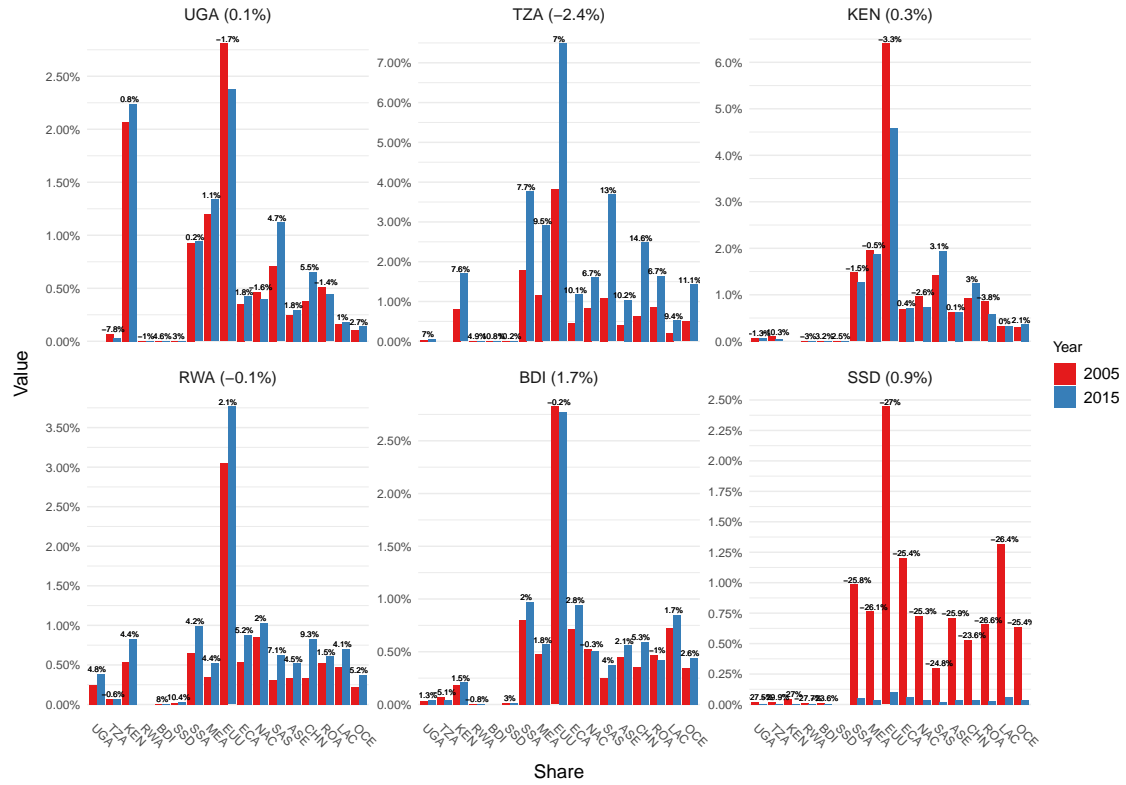
to their production, at overall higher foreign content shares of around 29% foreign VA content in Tanzanian production by 2015, and around 15% in Kenya. Kenya adds around 1-2% to Tanzanian production, and Uganda adds about 0.25% to Kenyan produce. The value added share of Kenya and Uganda in Rwandan production of around 1% and 0.5%, respectively is also visible. It is curious to observe the VS appears to have declined in Uganda, Kenya, Randa and Burundi from 2011 onwards, whereas in Tanzania it appears to have increased.

Figure 12: FOREIGN VALUE ADDED SHARES IN EAC PRODUCTION (VS)



To better summarize the movements in VA shares by different origins over the analyzed period, Figure 13 shows bars giving the value added share in 2005 and in 2015, and above the two bars the annualized average growth rate in the share over this period. The annualized average growth rate of the domestic value added share is also reported in round brackets behind the country code for each EAC country. Figure 13 makes it clear that the foreign value added share has increased substantially in Tanzania and also in Rwanda, but decreased slightly in Kenya and Uganda between 2005 and 2015. Examining more closely the inner EAC shares, it appears that Tanzania's VA share in Ugandan production has halved between 2005 and 2015, whereas Kenya's share has increased by 0.8% annually. For Tanzania the VA share of Kenya has almost doubled from about 0.8% to 1.8%. Rwanda has also seen increases in the shares of Ugandan and Kenyan VA of about 5% each year. In Burundi it appears foreign VA declined by 1.7% over the sample period with only minor changes in the shares of different suppliers. The data for South Sudan is considered unreliable, also minding that the country only became independent in 2011.

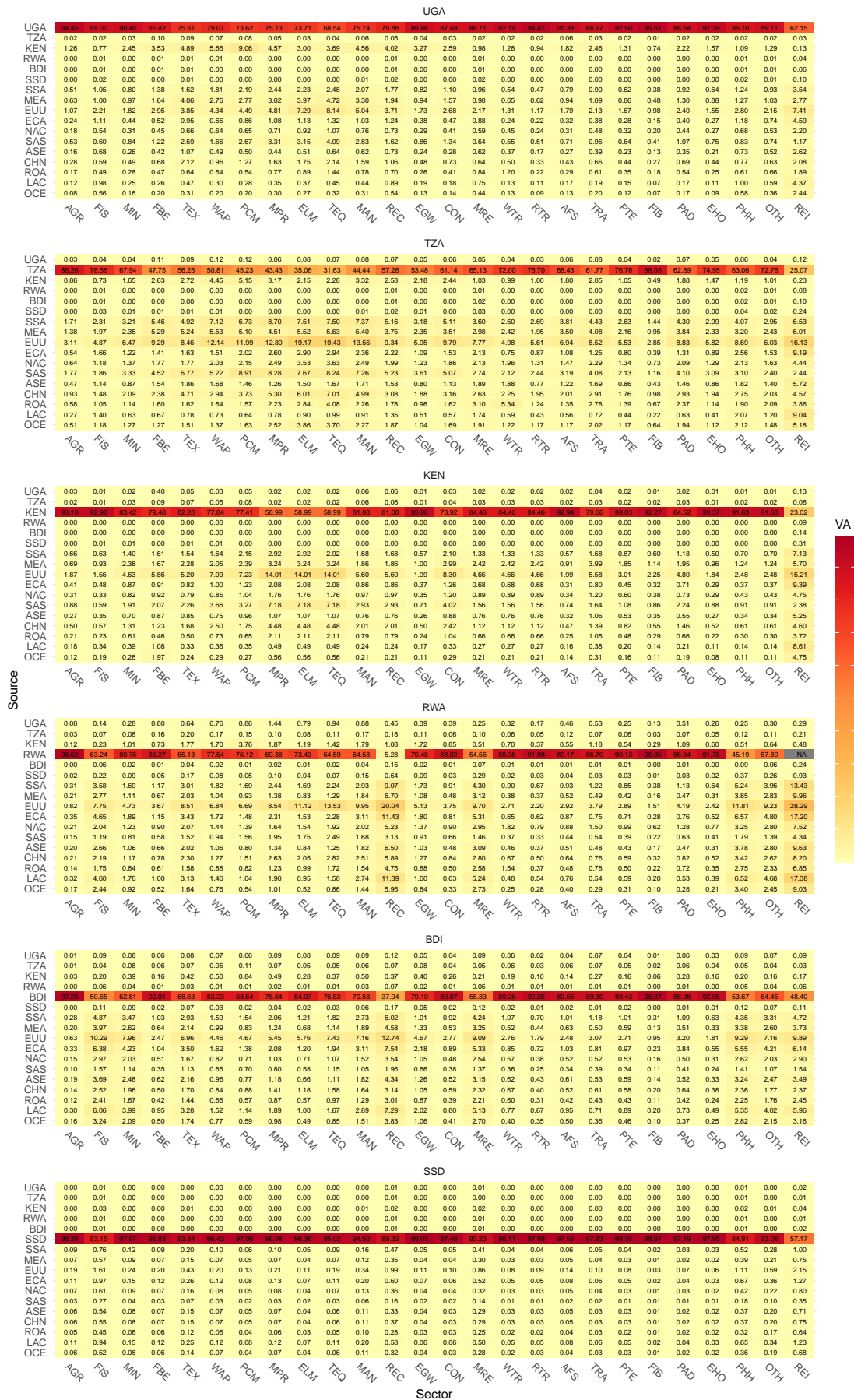
Figure 13: CHANGE IN FOREIGN VALUE ADDED SHARES IN EAC PRODUCTION



A full sector level view of **VB** for the EAC is not very informative due to the prevalence of very low shares. Figure 14 gives a partial sector disaggregation showing the value added in each EAC sector from different EAC countries and world regions. Across the different countries the domestic value added shares are lowest in the manufacturing sectors, coming as low as 30-35% in Tanzanian electrical machinery and transport equipment. Kenya is an important supplier of inputs and therefore of VA to these manufactured outputs, accounting for as much as 9% of the VA in Ugandan petro-chemicals. It was mentioned above that Uganda supplies a significant part of its Agricultural exports to Kenyan food and beverage industries. In VA terms Figure 14 suggests these inputs nevertheless only account for 0.5% of the VA in this industry in Kenya, which is likely explained by the high amount of VA in processing as well as by other inputs. Uganda has a greater role in Rwandan production, accounting for around 0.5-1% of the value added in several Rwandan manufacturing sectors.

In terms of other world regions, it is evident that the EU has significant shares in the produce of EAC manufacturing sectors, accounting for as much as 20% of VA in Tanzanian electrical machinery and transport equipment, and 14% of VA in Kenyan metal products, electrical machinery and transport equipment. In Uganda the EU shares are lower at 7.3% for electrical machinery and 8.1% for transport equipment. It is also evident that Rwanda, and to a lesser extent Burundi maintain highly international recycling industries, with significant inputs from all world regions. Other significant contributions to VA in EAC manufacturing come from South Asia, China and the rest of Sub-Saharan Africa, with shares ranging between 2 and 8%, depending on the country and sector.

Figure 14: DISAGGREGATED VALUE ADDED SHARE TABLES: EAC IN 2015
Shares in Percentage Terms, Columns Sum to 100 Percent



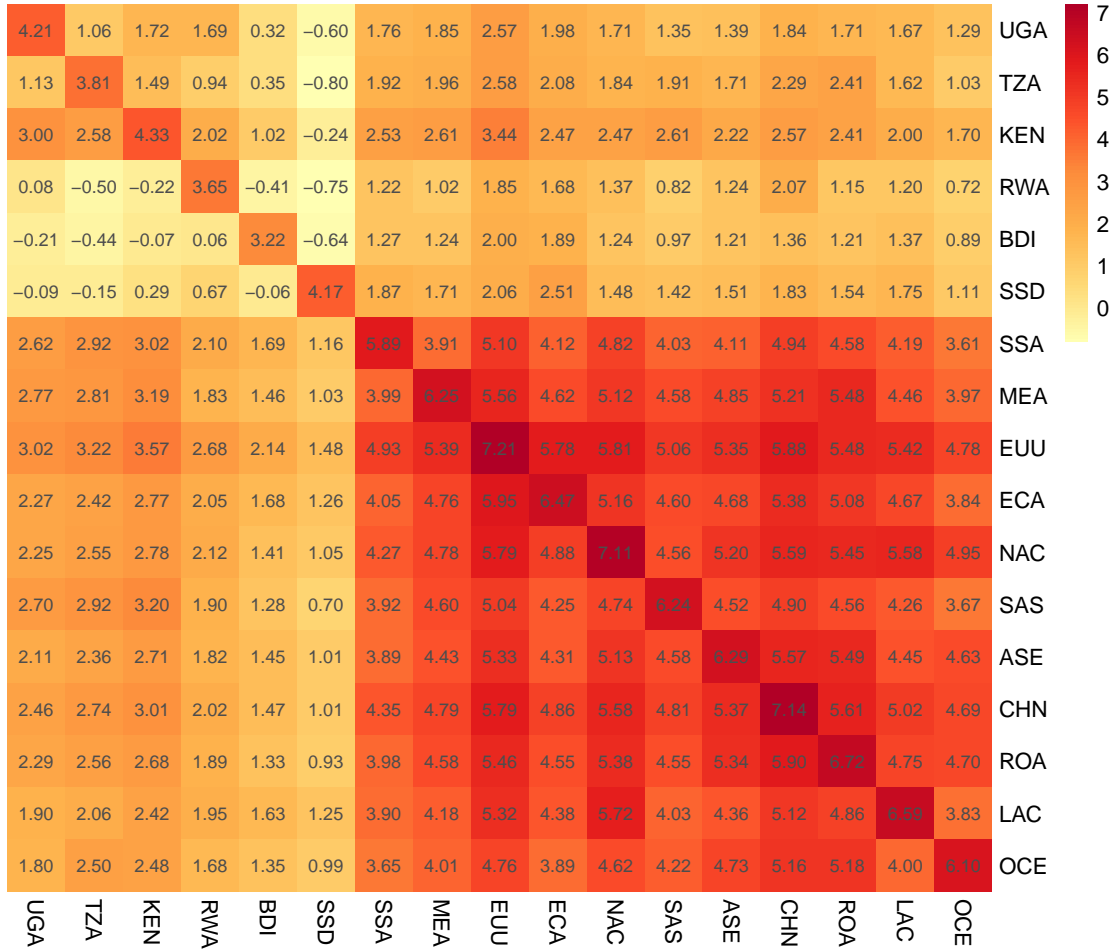
4.3 Intermediate Inputs in Value Added Terms

A first exercise could be to use **VB** to decompose the transaction matrix **T** of gross IO flows into VA terms, to see who adds value in who's production.

$$\mathbf{T}^{VA} = \mathbf{VBT}. \quad (8)$$

Figure 15 shows the transaction matrix in VA terms, analogous to the gross flows matrix shown in Figure 4. On first sight it appears that the EAC production shares are slightly smaller, with more VA accounted for by the rest of the World than EAC neighbours. Note that shares computed on this matrix differ from the **VB** shares since **VB** only concerns the production of one unit of the final good in each country-sector whereas **VBT** decomposes all flows feeding into heterogeneous amounts of final goods produced in different country-sectors. Compared to **VB** (Figure 11), the foreign VA shares in EAC countries derived from Figure 15 are around 10 percentage points higher.

Figure 15: AGGREGATED MRIO TABLE IN VA TERMS: EAC AND WORLD REGIONS
Millions of 2015 USD at Basic Prices on a Log10 Scale



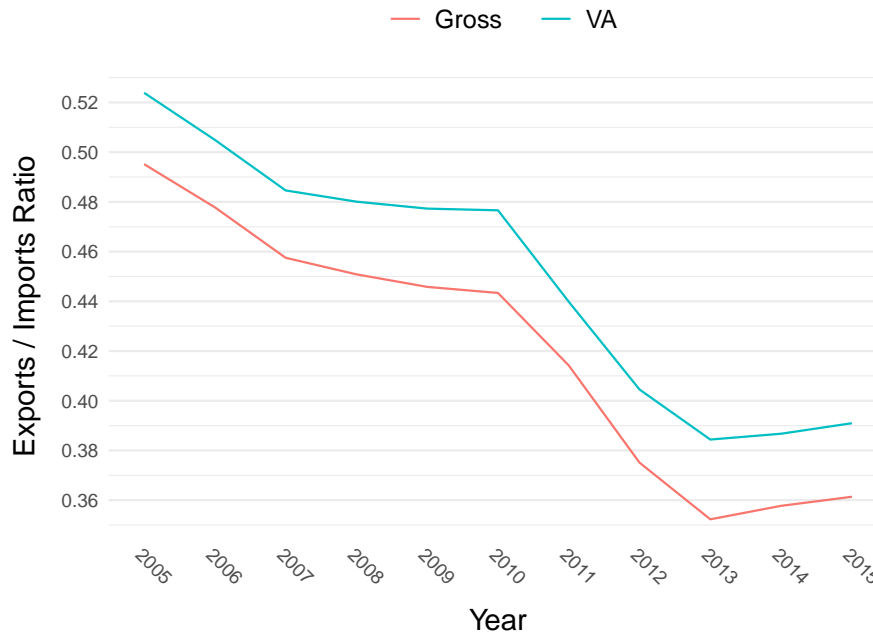
Tables 5 and 6 show the largest flows involving EAC countries in VA terms, analogous to Tables 3 and 4 recording gross flows. The ranking of the flows by magnitude is broadly similar in gross and VA terms, but it is evident in Tables 5 that Kenyan agricultural inputs to EU food processing industries are 30% VA larger in VA terms, whereas EU inputs to Kenya, for example in Transport, are around 15% smaller. This appears to give Kenya a favorable position in VA trade with the EU, at least in terms of the main traded intermediate goods. A similar pattern can be observed on the right side of Table 5 where Kenya is excluded: incoming flows to EAC countries are between 20% and 50% smaller in VA terms whereas EAC countgoing flows are 10-30% larger.

Table 5: LARGEST INTERMEDIATE FLOWS BETWEEN THE EAC AND THE WORLD IN VA TERMS
Millions of 2015 USD at Basic Prices

#	Flow	Value	Non-Kenya Flow	Value
1	KEN.AGR → EUU.FBE	597.299	EUU.ELM → TZA.ELM	86.850
2	KEN.AGR → EUU.REI	287.130	EUU.ELM → UGA.ELM	63.770
3	MEA.TRA → KEN.TRA	156.711	TZA.AGR → ROA.FBE	61.865
4	EUU.TRA → KEN.TRA	151.987	UGA.AGR → EUU.FBE	53.974
5	EUU.FIB → KEN.CON	120.347	EUU.FIB → TZA.FIB	49.740
6	EUU.ELM → KEN.CON	119.281	EUU.FIB → TZA.ELM	45.964
7	KEN.AGR → EUU.AGR	109.486	EUU.PCM → TZA.PCM	44.280
8	EUU.PCM → KEN.CON	100.395	TZA.AGR → EUU.FBE	43.597
9	OCE.AGR → KEN.FBE	94.364	MEA.ELM → UGA.ELM	38.484
10	EUU.PCM → KEN.PCM	94.076	EUU.ELM → TZA.TEQ	35.596
11	EUU.FIB → KEN.TRA	93.883	EUU.FIB → UGA.ELM	33.491
12	KEN.AGR → EUU.AFS	91.138	EUU.FIB → UGA.FIB	32.685
13	EUU.ELM → TZA.ELM	86.850	EUU.ELM → TZA.FIB	31.445
14	EUU.FIB → KEN.FBE	83.665	SSA.FIB → TZA.FIB	31.237
15	EUU.FIB → KEN.PCM	79.533	EUU.FIB → TZA.EHO	30.858
16	EUU.PCM → KEN.FBE	76.573	EUU.FIB → TZA.PCM	29.350
17	KEN.FBE → EUU.FBE	76.269	ROA.WTR → TZA.WTR	29.330
18	KEN.AGR → EUU.PCM	74.543	SAS.PCM → TZA.PCM	29.056
19	EUU.ELM → KEN.ELM	73.309	SSA.ELM → TZA.ELM	28.566
20	EUU.PCM → KEN.AGR	72.236	EUU.MPR → TZA.ELM	28.528

The discrepancy between gross flows and VA flows observed here is mostly explained by the nature of products traded between the EAC and ROW: The EAC mostly exports agricultural and other primary products with high domestic VA, whereas the EU and others mostly export manufactured goods and other complex products to the EAC, the creation of which involves imports of intermediates, and thus lower domestic VA. The result is that gross IO flows understate the relative importance of the EAC as a supplier of inputs to ROW. This finding is summarized by Figure 16, which shows the trade balance in intermediate goods, expressed as the ratio of intermediate exports to intermediate imports, between the EAC and the rest of the world, in gross and VA terms. It is evident that the value of intermediates supplied to the EAC is more than twice

Figure 16: EAC TRADE BALANCE IN INTERMEDIATE GOODS IN GROSS AND VA TERMS



as large than the value of intermediates supplied by the EAC, and this ratio has been deteriorating between 2005 and 2013 so that gross intermediate imports in 2013 were 3 times larger than gross intermediate exports. It is however also evident in Figure 16 that this ratio in VA terms is consistently larger by a value of 0.03. For example in 2015 EAC intermediate exports were 36% of intermediate imports in gross terms, but 39% of intermediate imports in VA terms.

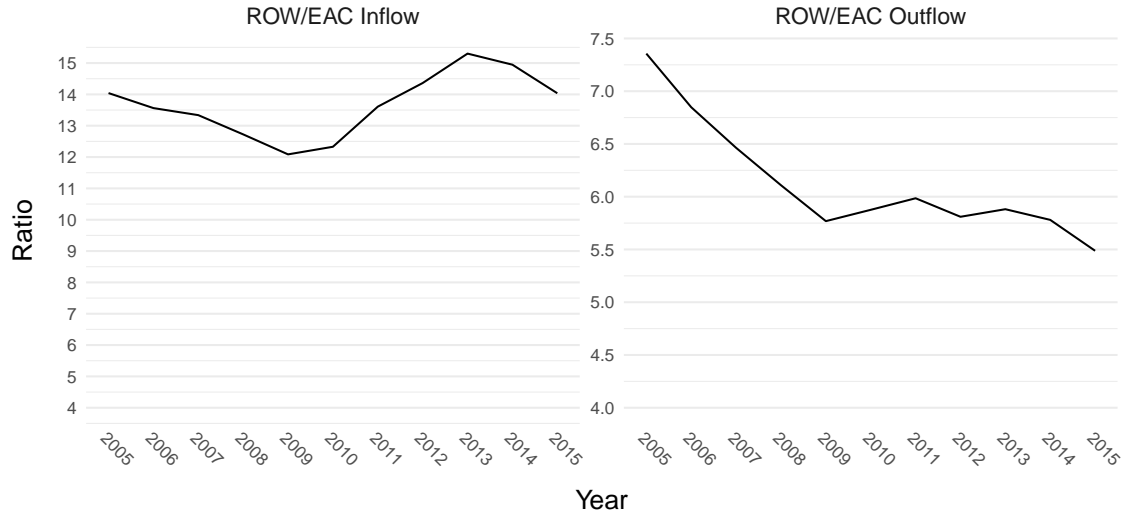
Table 6 shows the 20 largest intermediate flows within the EAC in VA terms. Compared to Table 4, these flows are smaller by 10-60%, which is due to the dominance of manufacturing inputs with lower domestic VA shares. In some cases the difference can be quite significant, for example inputs from the Ugandan to the Rwandan petro-chemical industry in 2015 were 2.54 million in gross terms, but 1.17 in VA terms. Primary flows increase in importance when considering VA terms, e.g. Ugandan agricultural inputs to Kenyan food processing industries are the 4th largest inner-EAC flow in VA terms, but only the 6th largest in gross terms.

Table 6: LARGEST INTER-COUNTRY INTERMEDIATE FLOWS WITHIN THE EAC IN VA TERMS
Millions of 2015 USD at Basic Prices

#	Flow	Value	Non-Kenya Flow	Value
1	KEN.MIN → UGA.PCM	80.599	UGA.TRA → RWA.PAD	1.720
2	KEN.PCM → UGA.PCM	45.927	UGA.TRA → RWA.TRA	1.366
3	KEN.PCM → TZA.PCM	23.150	UGA.AGR → RWA.FBE	1.215
4	UGA.AGR → KEN.FBE	22.839	UGA.MPR → RWA.MPR	1.181
5	KEN.WAP → UGA.WAP	20.393	UGA.PCM → RWA.PCM	1.170
6	KEN.ELM → UGA.ELM	19.860	UGA.MPR → RWA.ELM	1.109
7	KEN.PCM → UGA.EHO	18.594	UGA.FIB → RWA.ELM	1.054
8	KEN.FIB → UGA.PCM	16.246	UGA.FIB → RWA.PCM	1.007
9	KEN.MIN → TZA.PCM	14.535	UGA.FIB → RWA.PAD	0.885
10	KEN.MIN → UGA.EGW	14.426	UGA.FIB → RWA.FIB	0.851
11	KEN.PCM → UGA.CON	13.471	UGA.WTR → RWA.WTR	0.796
12	KEN.AGR → UGA.FBE	13.217	UGA.FIB → RWA.EHO	0.787
13	KEN.MIN → UGA.EHO	12.528	UGA.ELM → RWA.ELM	0.733
14	KEN.TRA → UGA.PAD	12.271	UGA.FIB → RWA.FBE	0.665
15	KEN.FIB → UGA.FIB	12.121	UGA.TRA → RWA.FIB	0.665
16	KEN.MIN → UGA.CON	11.871	TZA.MIN → UGA.PCM	0.659
17	KEN.PCM → UGA.FIB	11.832	TZA.FIB → UGA.FIB	0.658
18	KEN.WAP → UGA.CON	11.505	UGA.FIB → RWA.MPR	0.646
19	KEN.TRA → UGA.TRA	10.848	UGA.FIB → RWA.CON	0.645
20	KEN.MPR → UGA.ELM	10.620	UGA.FIB → RWA.TRA	0.579

Having found that both ROW inputs to the EAC and inner-EAC flows, being dominated by manufacturing inputs, are smaller in VA terms, it remains to assess their relative importance in VA terms. Figure 17 shows the ratio of ROW VA flows to the EAC to inner-EAC VA flows (excl.

Figure 17: VA FLOWS RATIOS: ROW/EAC INFLOWS AND OUTFLOWS



domestic flows), analogous to Figure 6. This ratio is higher in VA terms than in gross terms, with ROW providing on average 14 times more VA in EAC countries production than the EAC neighbours. In terms of outflows, a similar development as in Figure 6 is evident, with the relative importance of ROW declining from 7.4 times greater in 2005 to 5.5 times greater in 2015.

Thus in summary, although the EAC is more important to ROW in VA terms, as a supplier of (mostly primary) inputs, the relative magnitude of ROW inflows to the EAC compared to intermediate flows inside the EAC is greater in VA terms than in gross terms. Thus in VA terms trade in intermediates inside the EAC plays only a minor role compared to productive relationships between EAC members and ROW.

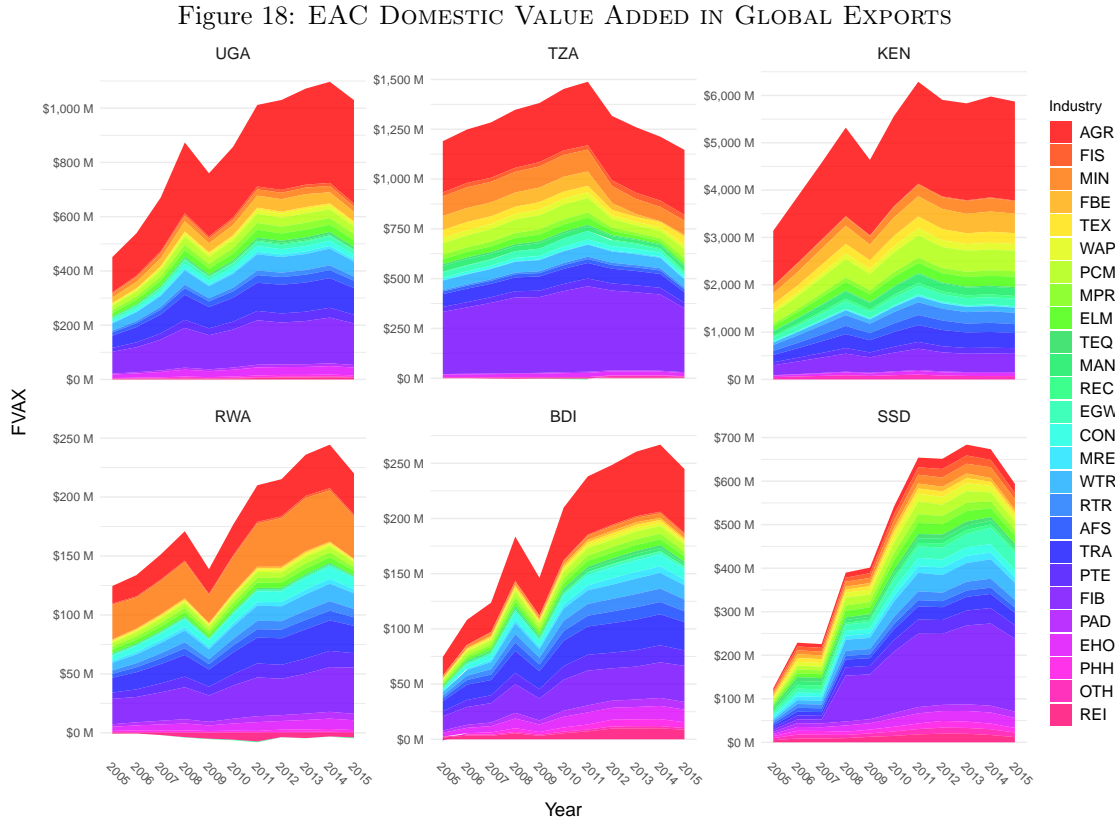
4.4 Value Added Exports

The matrix **VB**, reflecting the structure of international production, allows us to decompose any gross flow coming from any country-industry into its value-added origins by country-industry. The literature frequently considers a decomposition of gross exports into its VA origins, given by the matrix **VBE** where **E** is a diagonal matrix with exports by country-industry along the diagonal.

Figure 18 shows the EAC value added in global exports, which is obtained, for each country-industry, by simply summing the corresponding row of **VBE**.

$$VAE_{oi} = \sum_{uj} vbe_{oi,uj} \quad \forall oi. \quad (9)$$

Note that this therefore includes VA in exports to other EAC countries as well as VA in exports that are ultimately consumed at home. Figure 18 shows that the Ugandan VA in globally exported goods at the end of 2015 was around 1 billion USD in basic prices, whereas Kenya contributed around 6 billion USD to the value of global exports in that year.



Next to VS or the imported content of production and exports visualized in detail in Figure 12 which functions as a measure of backward GVC integration, [Hummels et al. \(2001\)](#) also introduced

the share of domestic exports that enters foreign countries exports, which they called VS1, as a measure of forward GVC Integration. This measure was first computed and explored by [Daudin et al. \(2011\)](#). It is defined as

$$VS1_{oi} = \frac{1}{E_{oi}} \sum_{uj, u \neq o} vbe_{oi,uj} \quad \forall oi, \quad (10)$$

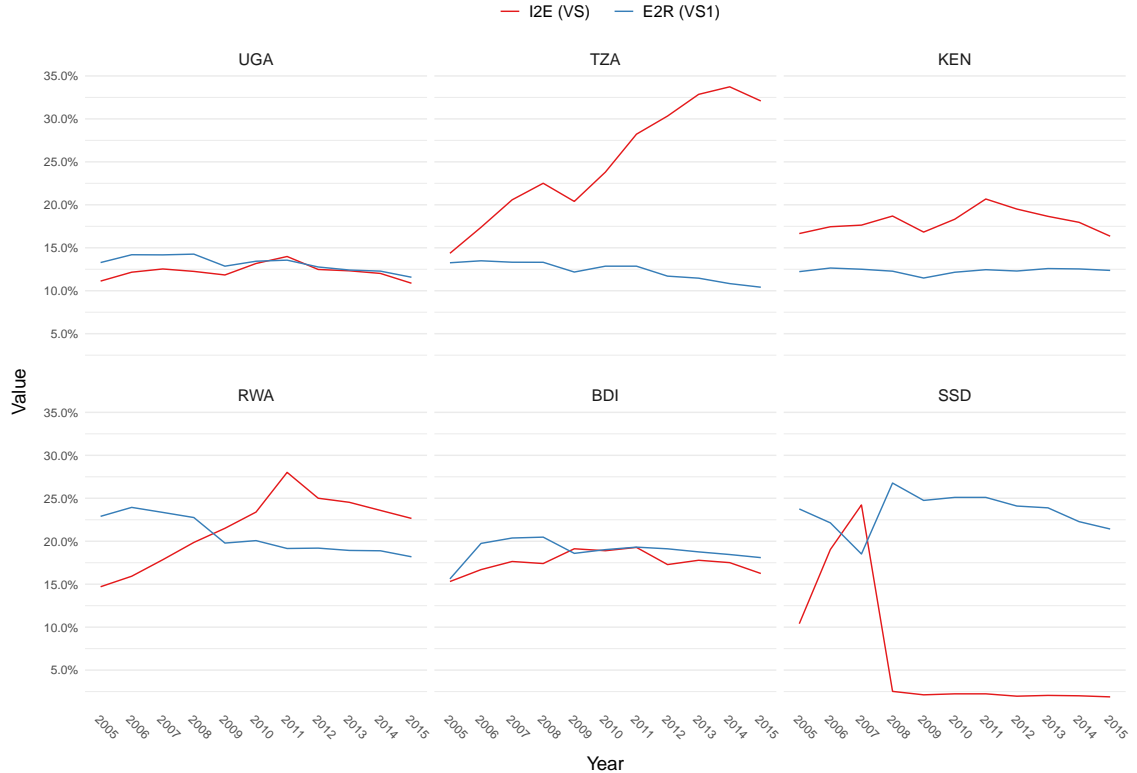
where E_{oi} are the gross exports of country-industry oi which is used to normalize the sum along the rows of **VBE** (excluding domestic industries) which capture the use of VA from a domestic sector oi in the exports of all foreign sectors uj . For completeness I note that VS can be defined in an analogous way as

$$VS_{uj} = \frac{1}{E_{uj}} \sum_{oi, o \neq u} vbe_{oi,uj} \quad \forall uj, \quad (11)$$

however since $\sum_{oi} vb_{oi,uj} = 1 \quad \forall uj$, the exports cancel out and Eq. 11 reduces to Eq. 7.

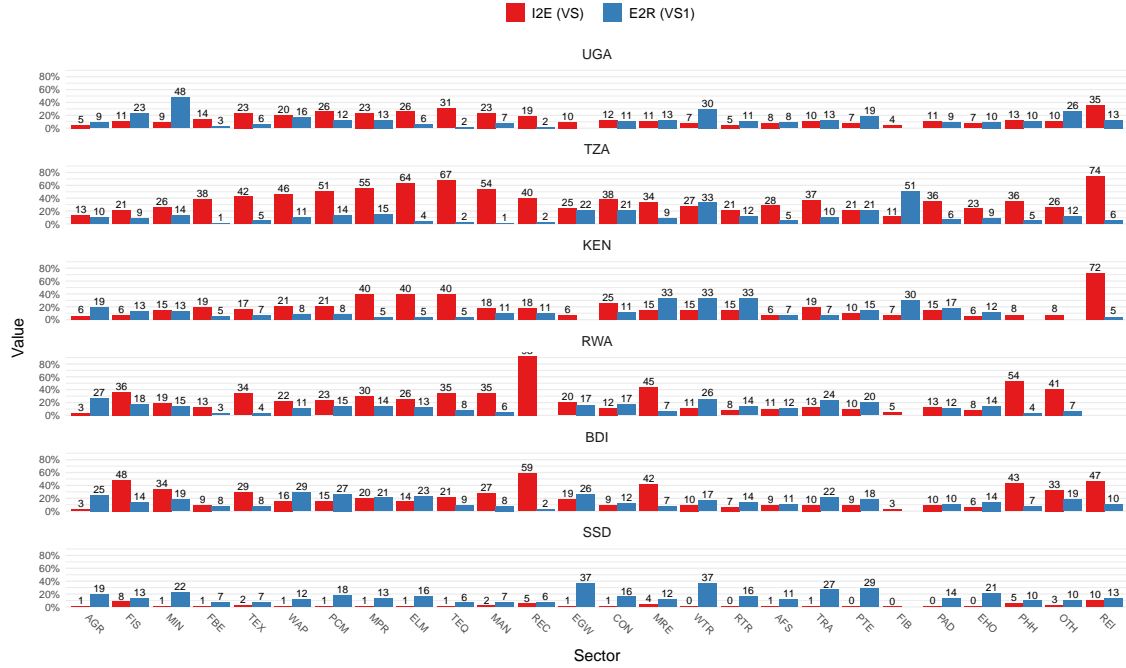
Figure 19 shows the aggregate VS and VS1 for each EAC member country over time. VS1 is called E2R (export to re-exports), and VS I2E (import to exports) by [Kummritz & Quast \(2016\)](#) (following [Baldwin & Lopez-Gonzalez \(2015\)](#)), which also developed the *gvc* R package to compute these measures. For Uganda, both VS and VS1 are at 11-12% of exports towards the end of 2015, whereas VS1 was higher at around 14% of exports in 2005. Tanzania, as already noted shows a remarkable increase in backward GVC participation to above 30% of VA in its produce generated abroad, but a slight decline in forward GVC participation down to 11% of exports being re-exported in 2015 - similar to Uganda. Kenya exhibits a stable development with VS of around 17% and VS1 around 12.5%. Rwanda increased in VS from 15% in 2005 to 22.5% in 2015, and at the same time showed a decrease in VS1 from initially 22.5% down to 16% in 2015. It is noteworthy that while a few members like Tanzania and Rwanda appear to have successfully increased their use of foreign intermediates, none of the members significantly increased it's role as supplier of inputs in the global market. Overall the GVC situation appears stagnant in Uganda, Kenya and Burundi.

Figure 19: GVC INTEGRATION OF EAC MEMBERS: AGGREGATE



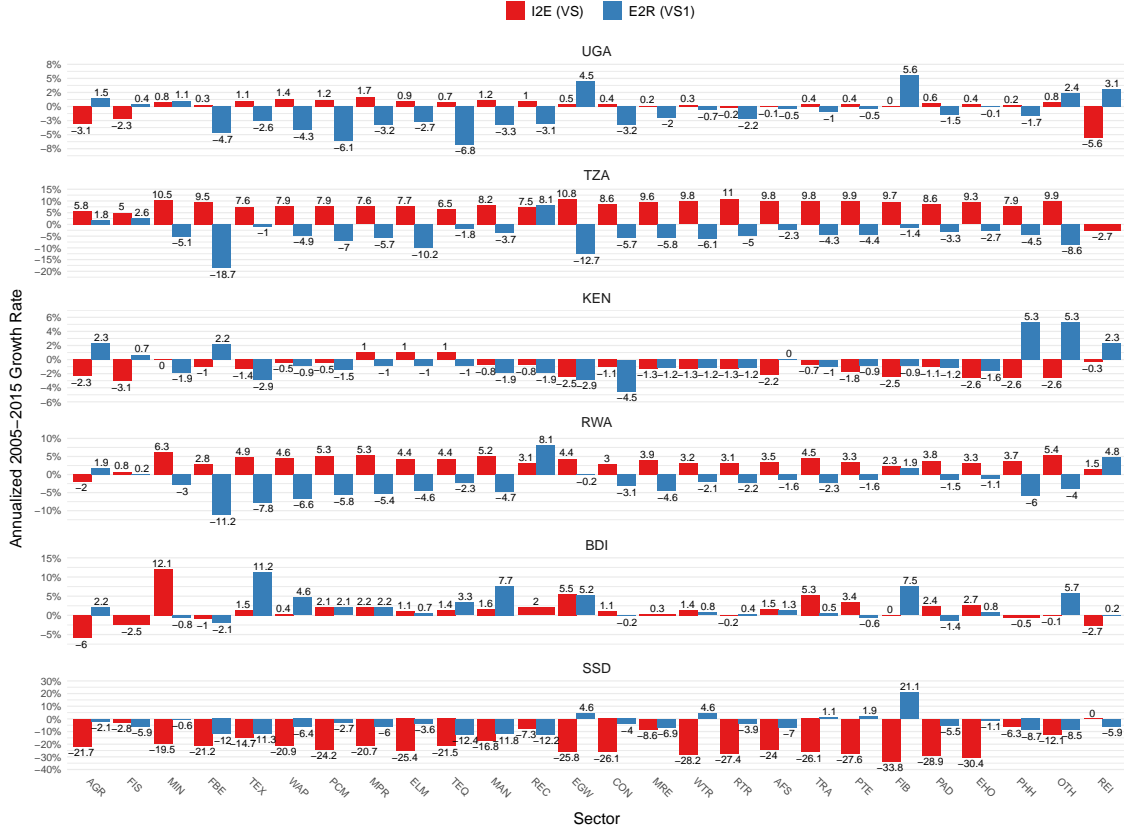
A sector-level snapshot of VS and VS1 for the year 2015 is provided in Figure 20. It shows the large share of foreign content in manufacturing, particularly in Tanzania. Rwanda and Burundi recycle a lot of foreign garbage. In terms of VS1, around 48% of Ugandan mining exports are manufactured into goods that are re-exported by receiving countries. Wholesale and retail trade also appears to have shares above 20% of exported goods being re-exported in all countries.

Figure 20: GVC INTEGRATION OF EAC MEMBERS: SECTOR LEVEL: 2015



To provide at least a rough overview of the sector-level dynamics over the last 10 years, Figure 21 shows the annualized average growth rates of VS and VS1. The axes are now on different scales for each country to reflect the large differences in sector level changes between countries. In Uganda and Kenya, developments are very moderate. Most Ugandan manufacturing sectors increased their imported content (VS) by 1-2% per annum, while at the same time reducing the share of intermediates for re-export (VS1) by 2-4% per annum. In Kenya the picture is more diverse, but with similarly small annual changes of 1-4%. A similarity between Uganda and Kenya is that both countries reduced their VS in Agriculture by around 3% per annum while increasing the share of exports for re-export (VS1) by 2%. In Tanzania nearly all manufacturing sectors exhibit annual gains in VS of 8-10%. I note again that the data reliability for Tanzania is very low and these estimates should be interpreted with great caution. Rwanda shows a similar development to Tanzania, but on a smaller scale. Here most manufacturing sectors appear to have increased their VS by 4-5% per annum, while reducing VS1 by a similar amount.

Figure 21: GVC INTEGRATION OF EAC MEMBERS: ANNUAL GROWTH 2005-2015



TODO: Do trade balance (ratios) in gross and VA terms for EAC members, and also state the relative importance of intermediate and final goods trade, both within EAC and between EAC and ROW.

4.5 Regional Integration in Value Added Trade

So far we have explored standard country and industry level metrics of backward and forward GVC integration, VS and VS1, and examined the overall level of GVC integration of EAC countries and sectors. The result of this examination was that apart from increases in the import content of production and exports (VS) in Tanzania and Rwanda, the situation is relatively stable over time, with moderate amounts of heterogeneity at the sector level.

This section builds on these findings and introduces some metrics to track regional EAC integration through VA in supply chains, relative to any global developments experienced by each member country. The first such metric computed is the share of foreign VA in a members production / exports accounted for by it's EAC partner states. It is computed as

$$VS_{uj}^{EAC} = \frac{1}{VS_{uj}} \sum_{oi \in EAC, oi \neq u} vb_{oi,uj} \quad \forall uj \in EAC, \quad (12)$$

where VS_{uj} is defined as in Eq. 7. VS_{uj}^{EAC} is thus a relative measure tracking the EAC share in VS, such that the overall EAC VA share in domestic production can be computed as $VS_{uj}^{EAC} \times VS_{uj} \quad \forall uj$. We can define an analogous measure for VS1 as the proportion of domestic VA in re-exported exports that is exported by EAC partner states.

$$VS1_{oi}^{EAC} = \sum_{uj \in EAC, u \neq o} vbe_{oi,uj} / \sum_{uj, u \neq o} vbe_{oi,uj} \quad \forall oi \in EAC. \quad (13)$$

These two metrics effectively track the role of the EAC in forming the interaction of each member country with the rest of the world in terms of production and export linkages. They do however not account for the import side, that is the overall role of the EAC in providing goods and services to each member country relative to the rest of the world. Therefore we will compute two additional metrics to capture this aspect of regional integration. The first metric is the share of EAC VA in the imports received by each member, which we shall denote by VAI^{EAC} . Consider E_u to be the vector of VA exports to EAC using country $u \in EAC$ from each country-sector⁴. Then we can find the VA origins of these exports to country u by pre-multiplying with \mathbf{VB} to give

$$E_u^{VA} = \mathbf{VB}E_u, \quad (14)$$

where E_u^{VA} denotes the vector, with elements $e_{oi,u}^{VA}$, of VA supplied by each country-industry (oi) in these imports of country u . From E_u^{VA} the share of EAC VA is easily computed as

$$\text{VAI}_u^{EAC} = \sum_{oi \in EAC, o \neq u} e_{oi,u}^{VA} / \sum_{oi, o \neq u} e_{oi,u}^{VA}. \quad (15)$$

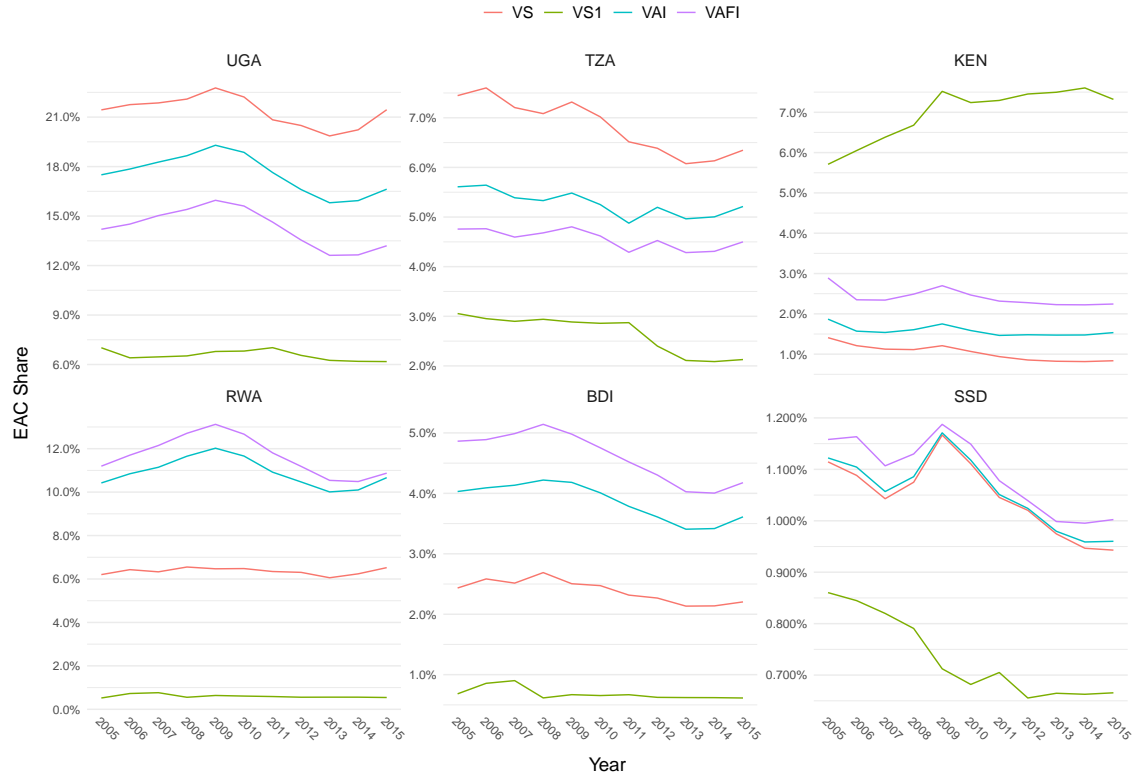
VAI_u^{EAC} thus gives us a country-level measure of the VA by it's EAC partners in it's import mix, excluding any domestic VA in imports. This VA may however enter into produced goods and be exported again, thus it is a measure of the EAC contribution to (non-domestic) production and consumption in a particular member country. To single out the EAC share in imported consumption goods, we need to consider only exports for final demand, which exclude exports feeding into production as intermediates. Let FE_u therefore denote the final exports to country u from each country-industry. Then $FE_u^{VA} = \mathbf{VB}FE_u$ denotes the decomposition of those exports into VA components, and we can define

$$\text{VAFI}_u^{EAC} = \sum_{oi \in EAC, o \neq u} f e_{oi,u}^{VA} / \sum_{oi, o \neq u} f e_{oi,u}^{VA} \quad (16)$$

as the EAC VA share in final goods exported to a particular member u .

⁴Note that since we don't have final demand disaggregated by sector, we cannot compute VAI^{EAC} by receiving sector, but only by receiving country.

Figure 22: EAC VA SHARES IN MEMBERS VS, VS1, IMPORTS AND FINAL IMPORTS



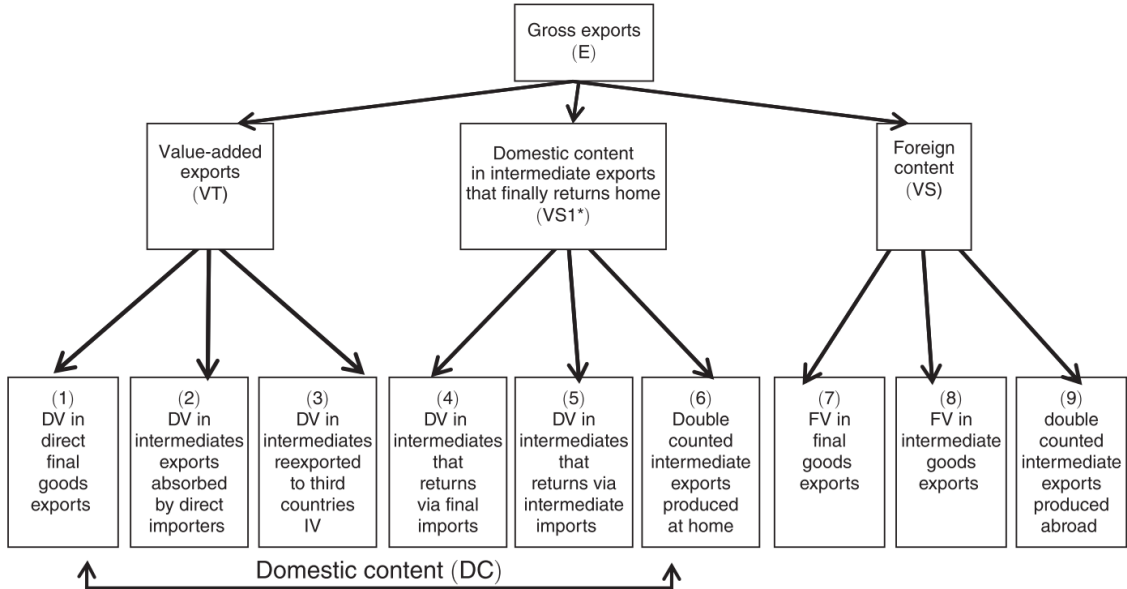
4.6 Koopman Wang Wei Decomposition of Gross Exports

One problem with the Leontief decomposition of gross exports into VA origins is that it also captures so called pure double counted items, which are items that are traded two or more times between the same trading partners. For example if on a value chain for chemical products an intermediate product would be first exported from Uganda to Kenya, processed further in Kenya and then imported again by Uganda to produce a final good that is the exported. The Leontief decomposition will correctly allocate the share of VA in this product to Uganda and Kenya, but Ugandan gross exports themselves would overstate the amount of VA in either of the two countries, because it includes both the export of the intermediate produce to Kenya, and the export of the final good to the rest of the world. So this kind of double counting in gross exports is incurred whenever there exists two-way trade in intermediate goods.

Secondly, the Leontief decomposition provides no information as to where and how the VA in exports is absorbed, it only provides the origin of VA in gross exports. To account for double counted items in gross exports and also to better understand where and how VA is absorbed, which indicates how countries integrate into GVCs, a number of increasingly complex GVC decompositions have been developed. The simplest and most well known of these is the Decomposition of country-level gross exports into 9 VA components proposed by Koopman, Wang and Wei (2014), henceforth KWW (Koopman et al., 2014). The 9 terms for the KWW decomposition are given schematically in Figure 23⁵.

The decomposition splits exports first into foreign content (VS) and domestic content. Domestic content is then further split into VA exports that are absorbed abroad, and content that eventually returns home and is absorbed domestically. Each of these three categories can be subdivided further according to how the VA is utilized. In both the domestic content returning home and foreign content, there are double counted categories, which split double counted VA arising from two-way trade in intermediate goods according to their domestic and foreign VA. In the example given above: if an intermediate is first exported from Uganda to Kenya, then re-imported by Uganda and finally exported, then the VA in that first export from Uganda to Kenya would be assigned to the domestic and foreign double counted terms (depending on where the value of the intermediate at that stage originated).

Figure 23: KWW DECOMPOSITION OF GROSS EXPORTS



⁵A mathematical expression for each of the 9 terms is provided in Eq. 36 of the Koopman et al. (2014) paper.

4.6.1 Aggregate KWW Decomposition

The KWW decomposition of gross exports is computed for each of the EAC members and shown in Figure 24. To connect this decomposition to the aggregate measures of GVC integration VS and VS1 obtained from the Leontief decomposition and shown in Figure 19: VS the share of FVA in gross exports is the sum of FVA_{FIN} (7), FVA_{INT} (8) and FDC (9), while VS1 is the sum of DVA_{INTrex} (3), RDV_{FIN} (4), RDV_{INT} (5) and DDC (6) (approximately equal to DVA_{INTrex} here since RDV_{FIN} , RDV_{INT} and DDC are close to 0 in all EAC countries).

Figure 24: KWW DECOMPOSITION OF GROSS EXPORTS

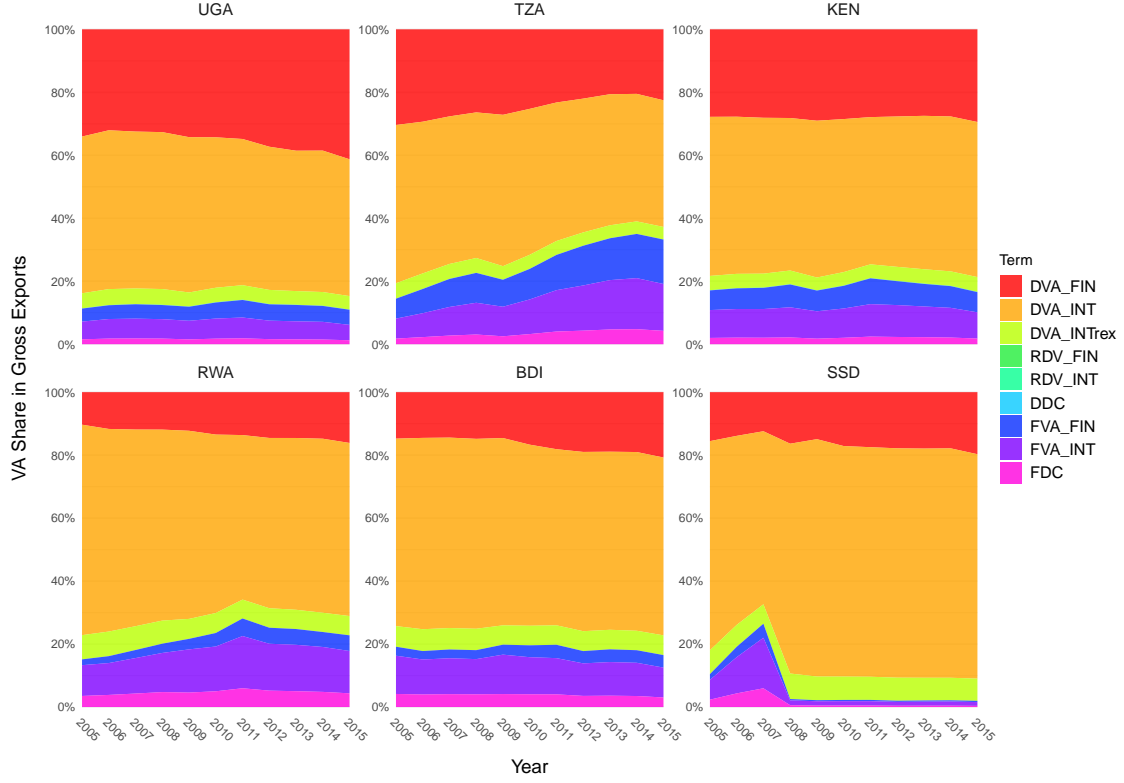


Figure 24 shows that double counted items constitute up to 10% of gross exports in EAC countries, but that most of this double counting occurs with VA produced abroad. The domestic content in intermediate exports that finally returns home is practically 0 for all EAC members, indicating an overall insignificant role of these countries as suppliers of inputs to their own final imports⁶. In all members furthermore the largest share of exports is domestic VA in intermediate exports absorbed by direct importers. Only a small share of DVA in intermediate exports is re-exported indicating that EAC countries predominantly export basic inputs to products manufactured for home consumption in the importing countries. In Uganda furthermore around 40% of exports constitute DVA in final goods exports. Apart from Tanzania, the share of DVA in final goods exports has increased over time. Tanzania is also the only country with a significant share of FVA in final goods exports, indicating some assembly and processing tasks. Rwanda on the other hand has a high share of FVA in intermediate exports⁷.

4.6.2 Upstreamness and Downstreamness in GVC Participation

The KWW decomposition also lets us assess the position of countries in GVCs regardless of their overall level of GVC integration⁸. According to Kummritz & Quast (2016) and Wang et al. (2013), High FVA in final exports relative to total foreign content in exports indicates downstreamness

⁶This is prevalent in the export composition of high-income countries, see e.g. (Kummritz & Quast, 2016).

⁷Primary the recycling sector as shown in Figure 28.

⁸As measured by VS and VS1.

(assembly tasks), while high DVA in intermediate exports relative to total DVA in exports indicates upstreamness (specialization in tasks adding a lot of value to an unfinished product). I follow these authors in computing the ratios as shown in Equations 17 and 18, and plot them in Figure 25.

$$\text{Upstreamness} = \frac{\text{DVA}_{INT} + \text{DVA}_{INTrex} + \text{DDC}}{\text{DVA}_{FIN} + \text{DVA}_{INT} + \text{DVA}_{INTrex} + \text{RDV}_{FIN} + \text{RDV}_{INT} + \text{DDC}} \quad (17)$$

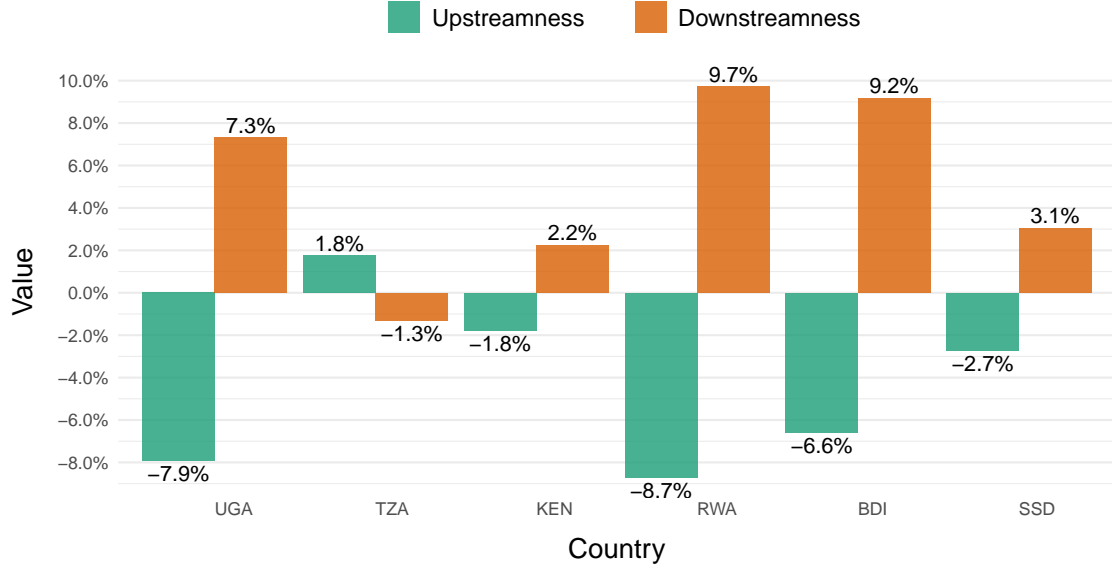
$$\text{Downstreamness} = \frac{\text{FVA}_{FIN}}{\text{FVA}_{FIN} + \text{FVA}_{INT} + \text{FDC}}, \quad (18)$$

Figure 25: UPSTREAMNESS AND DOWNSTREAMNESS RATIOS



From Figure 25 it appears that the smaller countries are situated more upstream in GVCs, but this is also a consequence of them generally exporting less final goods. More meaningful than the levels of these ratios is their change over time. Figure 25 suggests that apart from Tanzania all EAC members became more downstream in their GVC integration, with less domestic content in intermediate exports and more FVA in final goods exports. To better exhibit these findings, Figure 26 shows the difference in the Upstreamness and Downstreamness ratios between 2005 and 2015. It is evident that Uganda and Kenya moved downstream in these years, with more foreign value added going into final goods than intermediate goods, and also more DVA going into final goods. This could, especially for the smaller countries Rwanda and Burundi, also indicate a general increase in Final goods exports that has nothing to do with shifting patterns of GVC integration. Only Tanzania appears to have moved slightly upstream over this time period, with more domestic and foreign VA going into intermediate exports.

Figure 26: UPSTREAMNESS AND DOWNSTREAMNESS RATIOS, DIFFERENCE 2005-2015

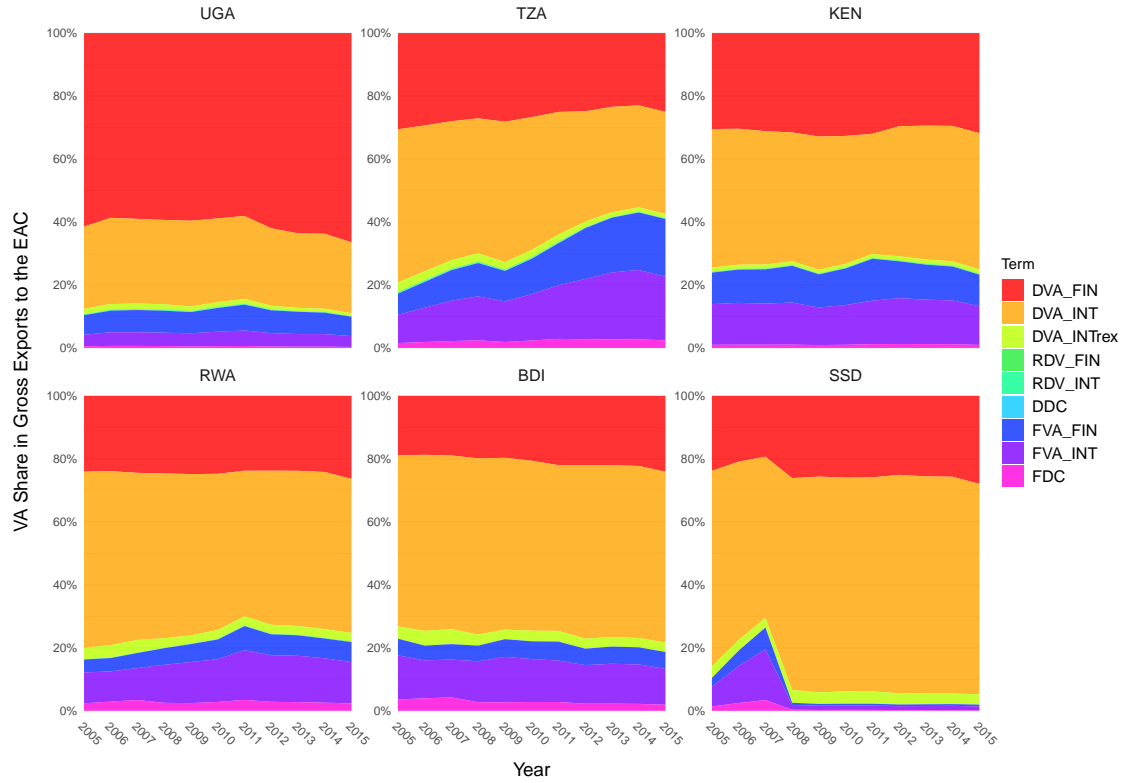


4.6.3 KWW Decomposition of EAC Exports to the EAC

Whereas the KWW decomposition is only defined at the country-level, Wang et al. (2013) derived a much more detailed decomposition that decomposes gross exports into 16 terms at the bilateral and sector-level. To consider 16 terms is a bit over the top given the very low data quality for developing countries. It is however possible to add up these 16 terms to get the 9 terms of KWW at a disaggregated level.

Figure 27 shows the KWW decomposition of gross exports to other EAC countries. Overall it is very similar to the overall export decomposition in Figure 24, but with a couple of noteworthy differences. The first is that in all EAC countries DVA_{FIN} has a higher and DVA_{INT} a lower share in exports to the EAC than in overall exports, implying that final goods exports are more prevalent in inner-EAC trade than in the overall trade mix of member countries. Secondly, $DVA_{INT_{rex}}$ and FDC components are both a lot lower in inner-EAC trade. Since both exports of intermediates and double counted components are features of GVCs, this therefore suggests that EAC member countries engage relatively more in GVCs with the rest of the World than with their EAC partners, or at least that inner-EAC value chains are shorter than other value chains EAC members engage in (Kummritz & Quast, 2016).

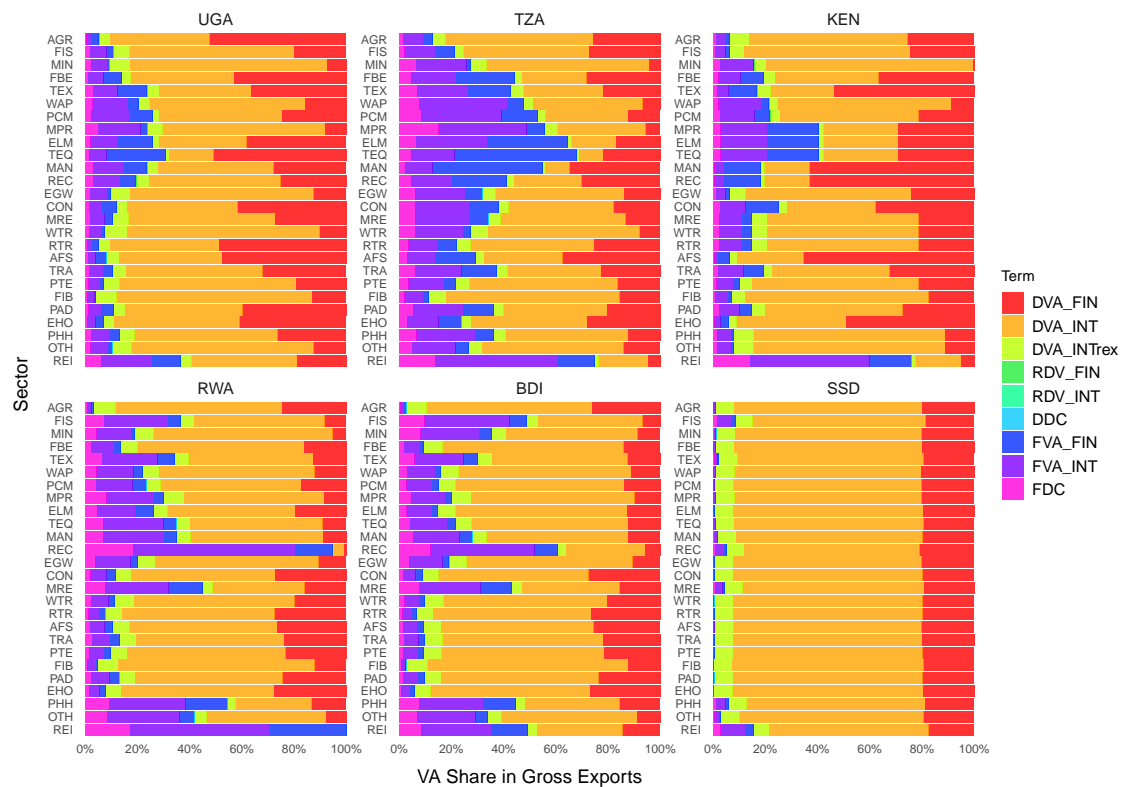
Figure 27: KWW DECOMPOSITION OF GROSS EXPORTS TO THE EAC



4.6.4 Sectoral-Level Decomposition

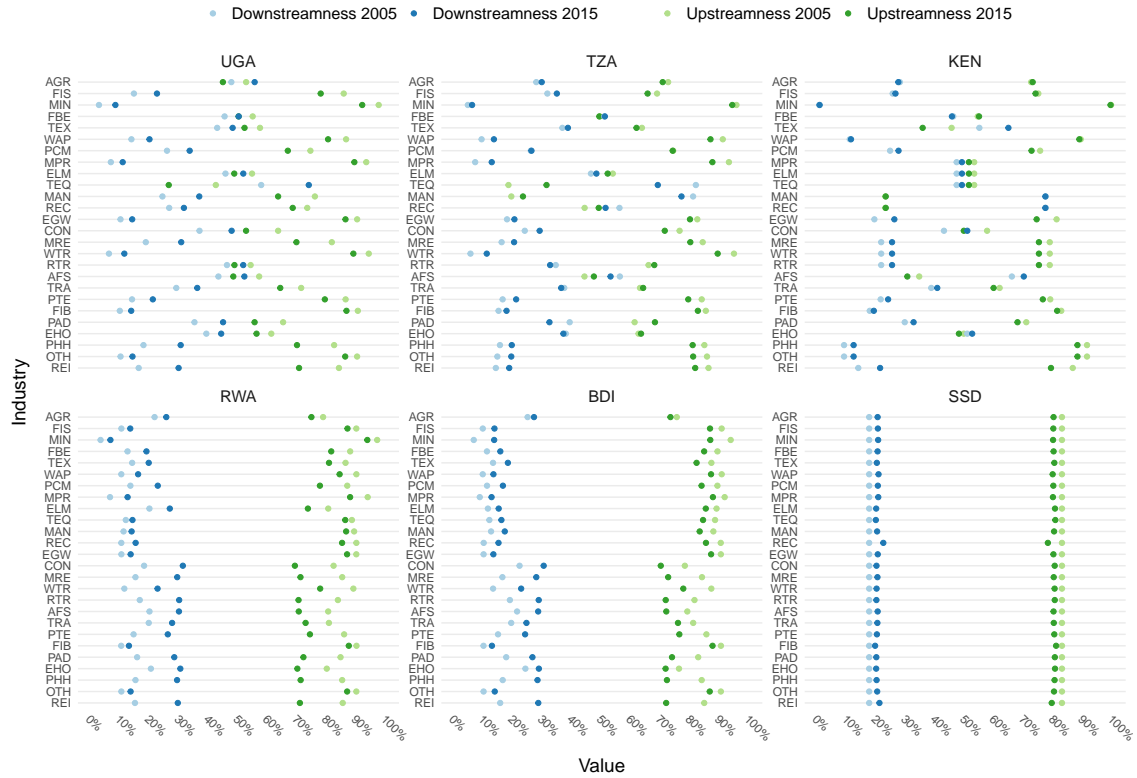
The decomposition of Wang et al. (2013) decomposes trade flows at the sector level, from which the sector-level KWW decomposition is easily obtained. Figure 28 shows this sectoral KWW decomposition for the year 2015. In general, manufacturing sectors have higher shares of FVA, including FDC, in all EAC countries. In some cases such as Rwanda's highly international recycling sector FVA is over 90%, where compared to Figure 20, Figure 28 provides the additional information that most of this recycling is re-exported as intermediate inputs.

Figure 28: KWW DECOMPOSITION OF SECTOR-LEVEL GROSS EXPORTS IN 2015



While Figure 20 already gives an indication of the overall share of FVA in sectoral exports for the year 2015, the more detailed decomposition in Figure 28 permits further investigation regarding the relative share of final goods vs. intermediates in both domestic and foreign VA. This information is difficult to read off Figure 28 but summarized by the Upstreamness and Downstreamness ratios computed above at the aggregate level, which Figure 29 now computes at the sector-level. To also gauge whether sectors have moved upstream or not, Figure 29 shows the values of Upstreamness and Downstreamness for both the first and last year of the analysis period.

Figure 29: UPSTREAMNESS AND DOWNSTREAMNESS BY SECTOR, 2005 AND 2015



While there is a great deal of sectoral heterogeneity in Figure 29, two findings stand out. The first is that in Rwanda and Burundi manufacturing sectors appear to be relatively more upstream than other sectors, which is not generally the case in Uganda, Tanzania and Kenya. The second more striking fact is that nearly all sectors in all EAC countries (including Tanzania) have moved downstream, that is to say exports have shifted more towards final goods than intermediates. In this regard movements at the sector level are very homogeneous and in line with the aggregate movement towards final goods production conveyed by Figures 25 and 26. This downstream shift will likely not increase the benefits EAC member countries can gain from participation in GVCs, as most of the technology transfer is likely to happen in upstream production tasks than in downstream assembly tasks. More research needs to be done however with regards to the nature and causes of this downstream shift before jumping to conclusions about its economic implications.

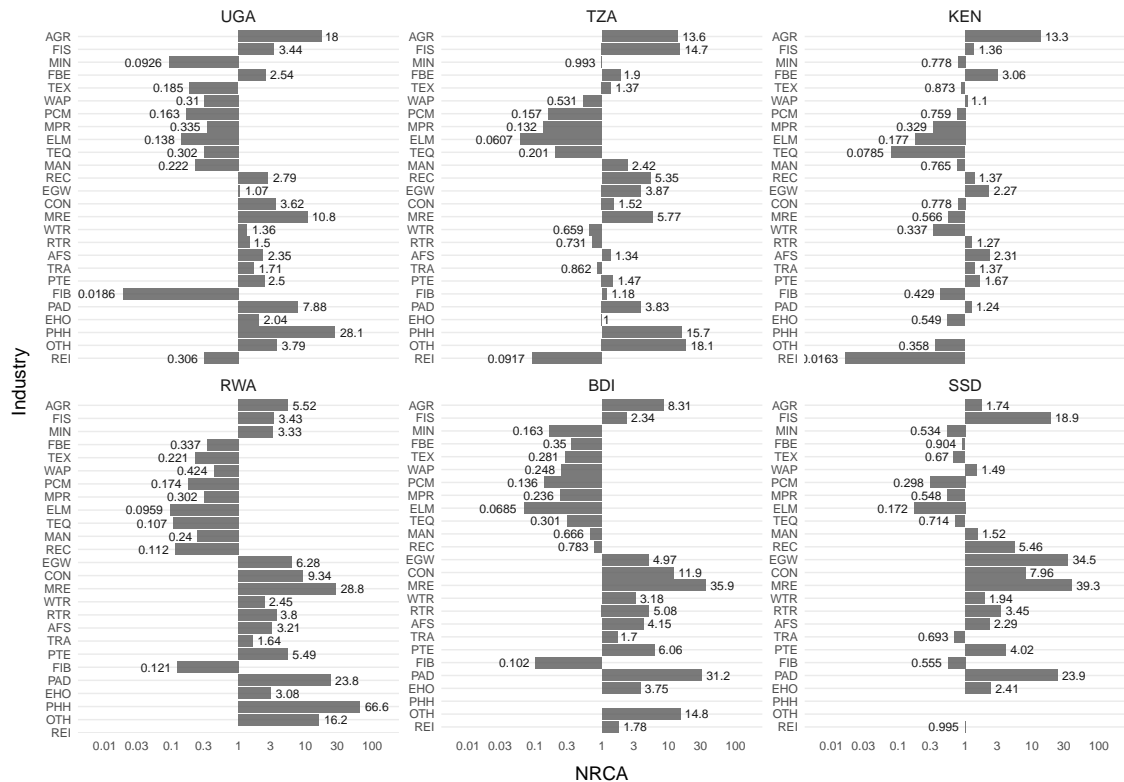
4.7 New Revealed Comparative Advantage

A popular measure to empirically measure Ricardo's concept of comparative advantage in international trade is the measure of revealed comparative advantage proposed by Balassa (1965). It is computed as the share of a sector in gross country exports, divided by the share that of that sector in gross World exports. A ratio above 1 indicates a comparative advantage of the country in this sector. The traditional index based on gross flows however does not take account of double counting in gross exports, and may thus be noisy and misleading. Koopman et al. (2014) therefore propose a new index based on VA flows, which considers the domestic VA in gross exports (or domestic GDP in exports, the sum of terms 1-5 of the KWW decomposition), to compute a new

revealed comparative advantage (NRCA) index.

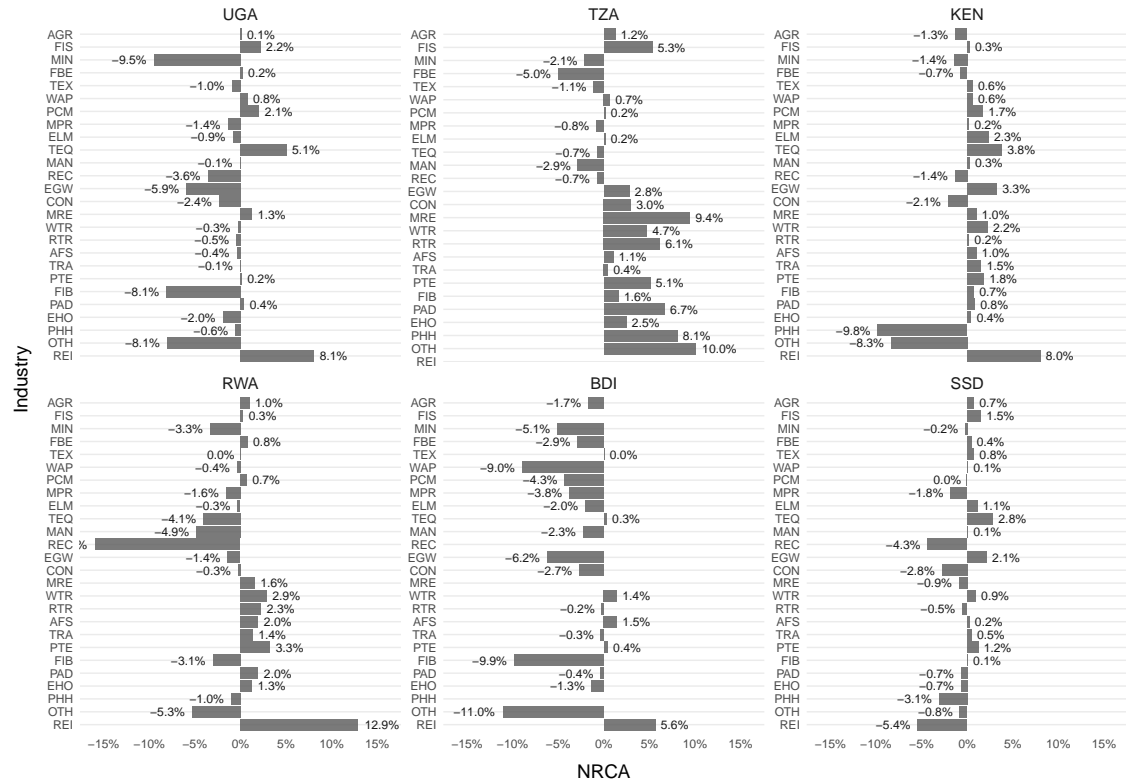
Figure 30 shows the NRCA for EAC countries in the year 2015. It is evident that all EAC members have a NRCA in agriculture and fishing, which is higher than 10 for agriculture in Uganda, Tanzania and Kenya. Also all EAC members have a comparative disadvantage in core manufacturing sectors such a petro-chemicals, metral products and electrical machinery. The remaining sectors show more heterogeneity across EAC countries, where Kenya appears to be different from the other countries. In Uganda, Tanzania, Rwanda, Burundi and South Sudan activities of private households (self-employment) seem to have a strong comparative advantage, and also maintenance and repair activities have a strong comparative advantage, whereas in Kenya both sectors appear to have a comparative disadvantage.

Figure 30: NEW REVEALED COMPARATIVE ADVANTAGE IN 2015



As we have 10 years of data, it will be interesting to see whether there were significant changes in the revealed comparative advantage of sectors. Figure 31 therefore shows the annualized growth rate in NRCA over the 2005-2015 period. It is evident that comparative advantage has not changed much in agriculture, with minor annual gains or losses within the $[-2\%, 2\%]$ range. All EAC members seem to have gained comparative advantage in fishing, particularly Uganda and Tanzania with gains of 2.2% and 5.3%, respectively. Also all EAC members had lost comparative advantage in mining, especially Uganda. All EAC countries have gained comparative advantage in re-exporting goods. In other sectors developments are rather heterogeneous. Uganda for example appears to have gained comparative advantage in exporting transport equipment by around 5.1% annually, whereas Rwanda lost comparative advantage in the same sector by -4.1% annually. There were also hardly any movements of sectors from comparative disadvantage ($NRCA < 1$) to comparative advantage ($NRCA > 1$) in this time period, only Tanzania's mining sector moved to a slight comparative disadvantage of 0.99 in 2015, from 1.23 in 2005.

Figure 31: NRCA ANNUALIZED 2005-2015 GROWTH RATE

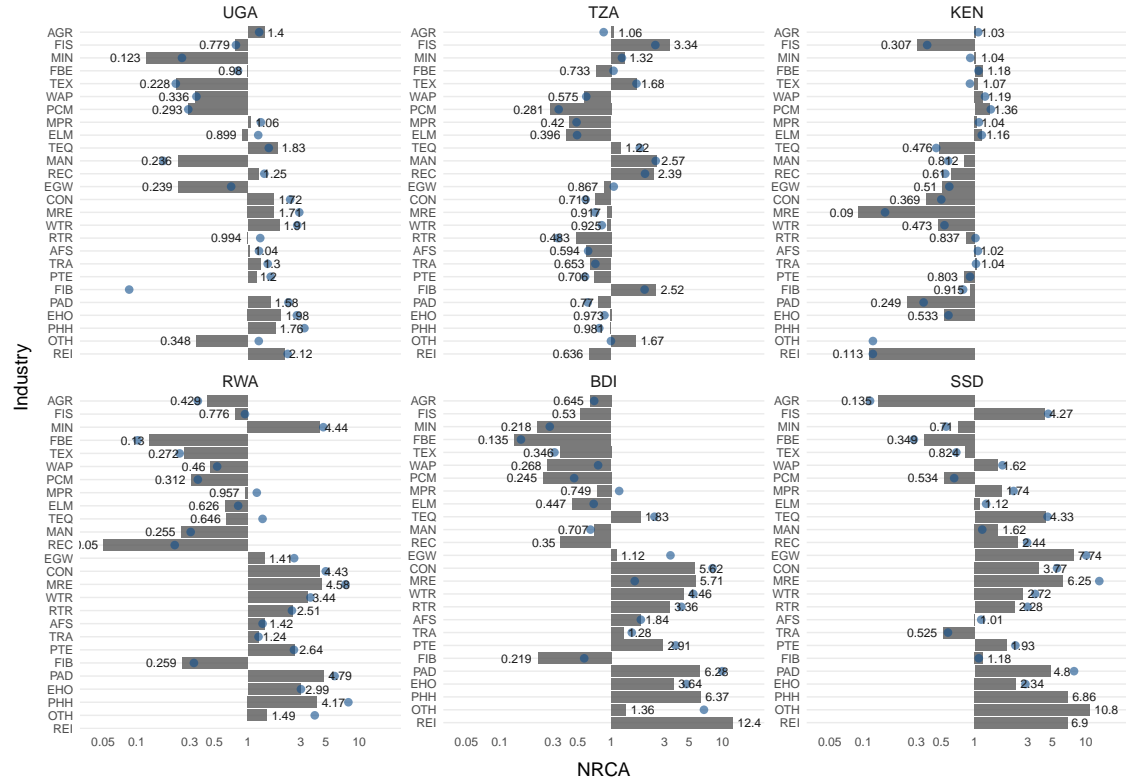


4.7.1 NRCA Relative to the EAC

Figures 30 and 31 have shown that relative to the rest of the world EAC members exhibit similar patterns of comparative advantage with a general advantage in agriculture and disadvantage in manufacturing. This could be constitutive to forming a common trade block with the rest of the world, supported by a currency union as is currently planned for 2024/25. Nevertheless comparing the EAC with the rest of the world may mask rivalries and shifts in comparative advantage between member countries. A final exercise in this section of the paper is thus to compute comparative advantage relative to the EAC as the share of a sector in country VA exports to the share of the sector in EAC VA exports. Figure 32 shows the results, where in addition to the bars for the year 2015, I have added a blue dot giving the 2005 value - to avoid having to draw two charts while still enabling a comparison over time as well.

It is evident from Figure 32 that relative to other EAC members Uganda has a comparative advantage in agriculture, Tanzania has a comparative advantage in fishing, Rwanda has a comparative advantage in mining, and Kenya has a comparative advantage in core manufacturing sectors such as wood and paper, petro-chemicals, metal products and electrical machinery. In addition, it appears that Rwanda and Burundi, and to a weaker extent Uganda, have a comparative advantage in construction, maintenance and repairs, wholesale and retail trade, whereas Tanzania appears to have a comparative advantage in other manufacturing, recycling, and financial and business services. Other sectors are more heterogeneous between countries and sectors. The afore mentioned comparative advantage relations appear to be stable since 2005, without major shifts within or across sectors between 2005 and 2015.

Figure 32: NRCA RELATIVE TO EAC

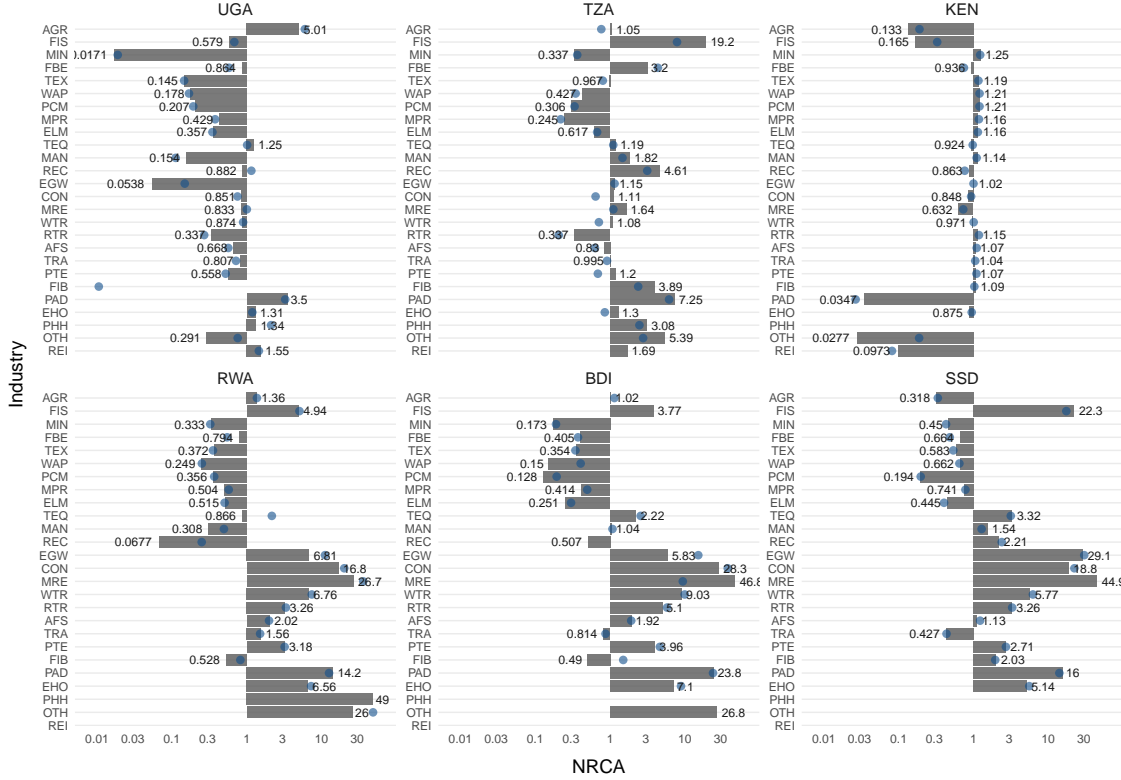


4.7.2 NRCA in Inner-EAC Trade

A final exercise that can be conducted around revealed comparative advantage in the EAC, is to compute NRCA only considering EAC trade with the EAC, and thus disregarding any trade flows between the EAC and the rest of the World. This can be useful to consider the evolution of trade relationships within the EAC itself, rather than questions of how the EAC as heterogenous trade block might engage with the rest of the World.

The results are shown in Figure 33, and are surprisingly similar to Figure 32. When considering only trade within the EAC, Uganda's comparative advantage in Agriculture and Tanzania's comparative advantage in fishing, recycling and financial and business services, as well as Rwanda's and Burundi's comparative advantage in construction, maintenance and repairs, wholesale and retail trade become more pronounced, whereas Kenya maintains a comparative advantage in core manufacturing sectors. An visible difference to Figure 32 is that Rwanda's comparative advantage in mining has disappeared and turned into a comparative disadvantage, indicating that the majority of Rwandese mining exports are to the outside World.

Figure 33: NRCA FOR INNER-EAC TRADE



5 GVCs and Industrial Development

Several papers have been written focussing on the global links between GVC integration and industrial development. As one of the first [Kummritz \(2016\)](#) assessed the role of GVCs for labour productivity and domestic value added using OECD ICIO's for 61 countries and 34 industries from 1995-2011. He achieves identification using a novel IV strategy where a value added trade resistance index combining third country trade costs with industry-specific technological variables induces exogenous variation in GVC participation. He shows that an increase in GVC participation leads to higher domestic value added and productivity for all countries independent of their income levels. His results imply that a 1 percent increase in backward GVC participation (I2E) leads to 0.11% higher domestic value added in the average industry, and a 1 percent increase in forward GVC participation (E2R) leads to 0.60% higher domestic value added and to 0.33% higher labour productivity ([Kummritz, 2016](#)).

The literature discussed by [Kummritz \(2016\)](#) outlines several channels through which GVC participation (in north-south value chains) increases the value added and productivity of its participants. The main channels are learning-by-doing, technology transfer or spillovers, gains from specialization in comparative advantage tasks, and terms of trade effects ([Kummritz, 2016](#)).

[Piermartini & Rubínová \(2014\)](#) for example use industry-level R&D and patent data for a sample of 29 countries during the period 2000-2008 and show that knowledge spillovers increase with the intensity of supply chains linkages between countries, and that these spillovers are larger in magnitude than spillovers from traditional trade flows. Similar evidence is presented by [Benz et al. \(2015\)](#) which use firm-level data to show that offshoring leads to knowledge spillovers, and further that forward spillovers (from producers to users if intermediate inputs) are stronger than backward spillovers.

[Kummritz \(2016\)](#) notes that GVCs however don't necessarily need to benefit developing countries as there could be adverse terms of trade effects or decreases in productive endowments from heavy engagement in them (which is also shown in some theoretical models such as [Baldwin &](#)

Robert-Nicoud (2014)). An argument made by Kummritz (2015) is also that GVCs might substitute foreign for domestic suppliers, but his own empirical research suggests that foreign value added works as a complement rather than a substitute to domestic value added and that GVC participation benefits the domestic economy along the value chain if certain prerequisites are met. The micro-papers discussed point towards an overwhelmingly positive effect.

Kummritz (2015) however finds no significant effect of GVC participation on low-income countries, which he attributes to the low absorptive capacity to benefit from technology spillovers.

Foster-McGregor et al. (2015) discusses industrial development in Africa through GVCs in the context of upgrading,

Four types of upgrading are often distinguished (Humphrey, 2004), the four types being: (i) process upgrading, which involves increased productivity in existing activities within a GVC; (ii) product upgrading, which is the movement into higher value-added products within a GVC; (iii) functional upgrading, which involves the movement into more technologically sophisticated or more integrated aspects of a production process; and (iv) inter-sectoral or chain upgrading, which involves a movement into higher value-added supply chains. The.

In their analysis Foster-McGregor et al. (2015) construct three alternative indicators, intended to capture one or more aspects of upgrading within GVCs.

-> They compare export unit values to export market shares The use trade data aggregated to EORA Sectors.

Compute Herfindahl Index of Value added exports and see how it varies with GVC involvement (Sum of VS and VS1 like in GVC in Africa Paper). Also compute oher upgrading and export difersification indicators. Take total GVC Involvement measure as in paper VS + VS1.

Combine GVC with your ES Analysis and look at innovtion measures.

Further empirical evidence on the relationship between domestic value chains i.e. the fragmentation of domestic production and GVCs is provided by Beverelli et al. (2019). They find that, across countries at different stages of development, higher domestic integration by 1 standard deviation raises subsequent GVC integration through backward linkages (I2E) by 0.4%. They also find that domestic value chain integration explains up to 30% of overall GVC integration. They explain these results with fixed costs of fragmentation and switching suppliers: "high fragmentation costs allow, due to their sunk nature, DVCs to act as stepping stones to GVCs" (Beverelli et al., 2019). The results also imply that improving domestic economic integration would further GVC integration in the medium run.

5.1 GVC Integration and Growth

A natural idea to assess the impact of GVC integraion on growth is to investigate if higher imported or re-exported content in exports is associated with higher domestic value added produced i.e. higher GDP. This can be examined, following papers such as Kummritz (2015), using a simple specification regressing the log of VA on the imported content share (I2E) and the re-exported content share (E2R). The specification is

$$\log(VA_{cst}) = \sum_{i=0}^p \beta_{1i} I2E_{cs,t-i} + \sum_{i=0}^p \beta_{2i} E2R_{cs,t-i} + \alpha_{cs} + \beta_{ct} + \gamma_{st} + \epsilon_{cst}, \quad (19)$$

where p is a suitable number of lags to include in the regression to allow changes in production (in particular I2E) to feed into greater productivity with a lag. The specification including lags is applied by Kummritz (2015) and theoretically justified by the dynamic GVC model of Li & Liu (2015) where the effect of GVC participation on domestic VA accrues in the next period. In the most general setting we can control for 3 sets of unobservable effects: country-sector effects (α_{cs}), country-year effects (β_{ct}) and sector-year effects (γ_{st}). A similar specification, without dynamics effects but with an instrument for I2E and E2R, is estimated by Kummritz (2016) who finds a positive effect of both GVC measures, and further that OLS and IV give similar results. This gives me some confidence that constructing an instrumental variable for GVC integration (which is difficult for these countries and the EORA data), might not add much to the estimation. Kummritz (2015) also notes that in the absence of an instrument, fixed effects and lags are the best specification

choices towards a careful causal interpretation of the results.

There is also the possibility to estimate a first-difference estimator which is more efficient if the autocorrelation in the error term is > 0.5 ⁹

$$\Delta \log(VA_{cst}) = \sum_{i=0}^p \beta_{1i} \Delta I2E_{cs,t-i} + \sum_{i=0}^p \beta_{2i} \Delta E2R_{cs,t-i} + \Delta \beta_{ct} + \Delta \gamma_{st} + \Delta \epsilon_{cst}. \quad (20)$$

For this regression I use data for Uganda, Tanzania, Kenya, Rwanda and Burundi. South Sudan was Excluded due to unreliable data. From the data for these countries I further remove sectors where $I2E$ or $E2R$ are greater or smaller than 1. This should usually not be the case but is the case in recycling and re-import/export sectors in Rwanda, Tanzania and Burundi, in the financial intermediation and business sectors in Burundi, Rwanda and Uganda, in Kenyan and Ugandan electricity gas and water, and Kenyan private households and other sectors. This is likely due to both bad data quality and very unusual economic activity in these sectors. For the estimation these six sectors (REC, REI, FIB, EGW, PHH and OTH in Table 2) are removed from the sample. Summary statistics for the excluded sectors are shown in Table 7.

Table 7: EXCLUDED SECTORS

Sector	GVC Measure	N	Mean	SD	Min	Max
EGW	I2E	55	0.143	0.055	0.063	0.248
EGW	E2R	55	0.750	0.539	0.158	1.828
FIB	I2E	55	0.059	0.024	0.027	0.110
FIB	E2R	55	4.603	5.225	0.294	19.275
OTH	I2E	55	0.224	0.122	0.077	0.493
OTH	E2R	55	3.164	6.194	0.072	21.512
PHH	I2E	55	0.307	0.179	0.077	0.658
PHH	E2R	55	3.071	6.240	-0.060	21.512
REC	I2E	55	0.435	0.258	0.173	1.018
REC	E2R	55	0.027	0.058	-0.108	0.137
REI	I2E	55	0.821	0.384	0.352	1.787
REI	E2R	55	0.028	0.100	-0.223	0.138

I end up with a balanced panel of $N = 1100$ observations in $CS = 100$ country-sectors (5 countries, 20 sectors) and $T = 11$ time periods. Summary statistics for overall variation as well as between and within country-sectors are shown in Table 8.

Table 8: SUMMARY STATISTICS OF VARIABLES

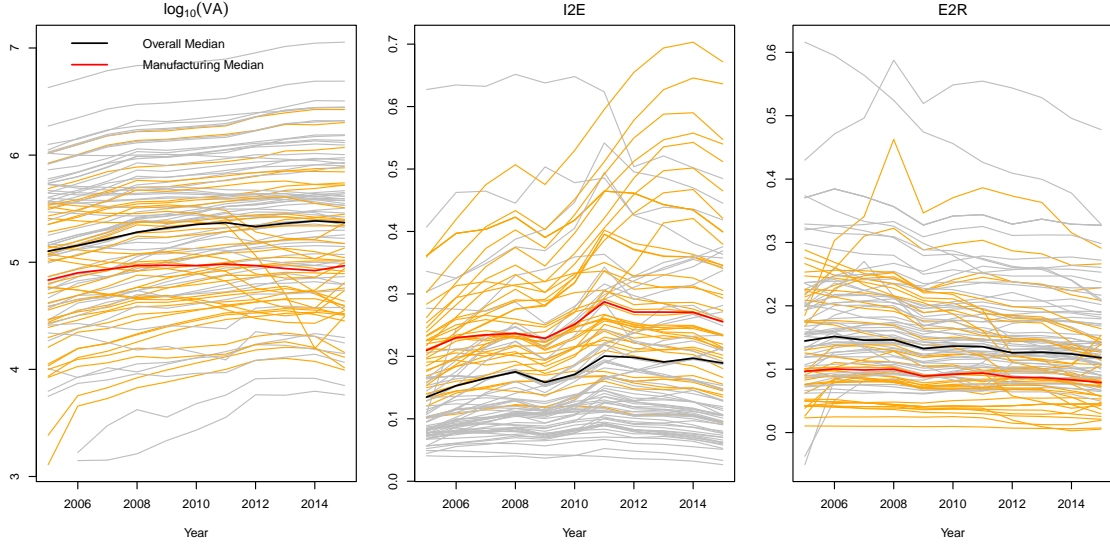
Variable	Trans.	N/T	Mean	SD	Min	Max
VA	Overall	1,100	499,353	993,836	-1,063	11,335,675
VA	Between	100	499,353	952,846	3,247	7,854,686
VA	Within	11	499,353	296,745	-3,084,848	3,980,341
I2E	Overall	1,100	0.20	0.13	0.03	0.70
I2E	Between	100	0.20	0.12	0.04	0.59
I2E	Within	11	0.20	0.04	0.01	0.36
E2R	Overall	1,100	0.15	0.10	-0.05	0.62
E2R	Between	100	0.15	0.09	0.01	0.51
E2R	Within	11	0.15	0.02	-0.00	0.30

To further expose the manufacturing sectors in these countries whose productivity is likely most affected by changing integration in GVCs, I also run regressions for a subsample of manufacturing

⁹In particular, if $\epsilon_{cst} = \rho \epsilon_{cs,t-1} + u_{cst}$, then $var(\epsilon_{cst}) = \rho^2 \sigma_\epsilon^2 + \sigma_u^2$ but for the first-differenced model $var(\Delta \epsilon_{cst}) = var(\epsilon_{cst} - \epsilon_{cs,t-1}) = var(\rho \epsilon_{cs,t-1} + u_{cst} - \epsilon_{cs,t-1}) = (\rho - 1)^2 \sigma_\epsilon^2 + \sigma_u^2$. So the FD estimator is more efficient if $(\rho - 1)^2 < \rho^2$ or if $\rho > 0.5$.

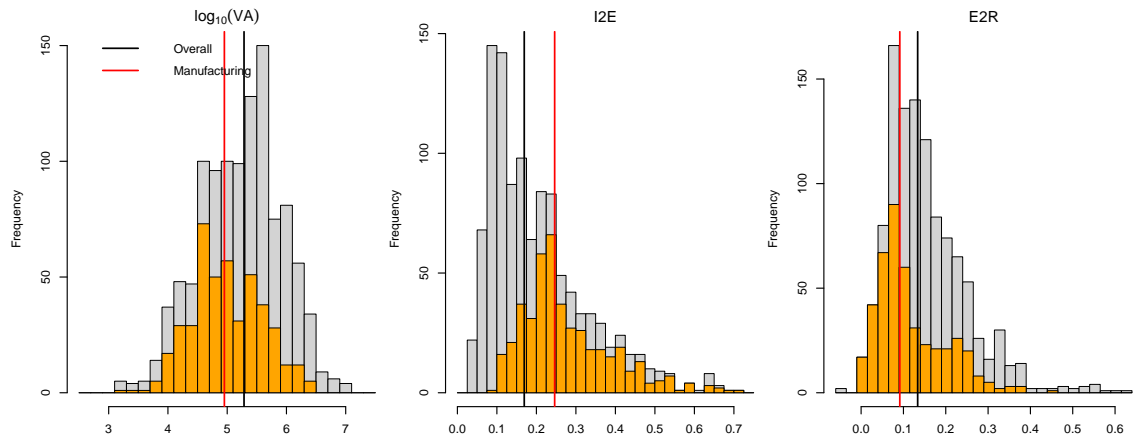
sectors (FBE, TEX, WAP, PCM, MPR, ELM, TEQ, MAN) in Table 2. To better understand the data and the manufacturing subsample of sectors, Figure 34 visualizes the data, where each orange line is a manufacturing sector in some EAC country, grey lines are other sectors, the red line represents the median value across all manufacturing sectors in a given year and the black line is the median across all sectors.

Figure 34: TIME SERIES OF VARIABLES



It is evident from Figure 34 that manufacturing sectors have a lower than average VA, a higher I2E share, and a lower E2R share, indicating that these sectors import more inputs that other economic sectors but export more final goods. The trend line is broadly parallel to the overall trend, but in the VA chart it appears that manufacturing sectors grew slower than the average. For a final step of visual investigation, 35 shows histograms conveying the same information as Figure 34. They show that in terms of VA manufacturing is a bit lower. In terms of I2E, manufacturing sectors make up the bulk of the upper part of the distribution, whereas for E2R the opposite is the case. In both cases the distribution is a bit skewed to the right.

Figure 35: HISTOGRAMS OF VARIABLES



Regarding the specification in Eq. 20, I first select the appropriate lag length by running the regression with fixed effects and first-differences and examining up to which order lags of I2E and E2R affect value added. Together with some judgement I opt for $p = 2$, which is a sensible choice as it might take up to 2 years for an innovation or change in supply chain to fully dissipate to output and productivity. Then, to determine whether the specification given in Eq. 20 is appropriate, I run a series of Hausman tests, including 2 lags of I2E and E2R on the regression. The first test evaluates the consistency of the random effects estimator against the the simple fixed effects esti-

mator with country-sector fixed effects, using the original χ^2 distributed quadratic form proposed by Hausman (1978). It rejects the null of random effects consistency $\chi_6^2 = 73.05$, $P < 0.01$. Then, I demean the data by country-sector and run a second Hausman test with country-year fixed effects. This test also rejects $\chi_6^2 = 53.4$, $P < 0.01$. Finally, I iteratively demean the data by country-sector and country-year until convergence and run a third Hausman test against sector-year fixed effects. This test also rejects $\chi_6^2 = 48.55$, $P < 0.01$, but a robust version of the test based on an auxiliary regression as specified in Wooldridge (2010)¹⁰ fails to reject the null $\chi_6^2 = 9.37$, $P = 0.15$. I nevertheless keep the sector-year fixed effects in the model, as the coefficient is practically identical to the one without them, and keeping them reduces a bit the serial correlation in the error term.

Serial correlation in the error term, $\epsilon_{cst} = \rho\epsilon_{cs,t-1} + u_{cst}$ for $\rho > 0$ might obscure conducting inference on the model and make the first-difference estimator more efficient. In practice it is difficult to determine the value of ρ , given that there are errors in the fixed effects model are unobservable¹¹, but I use $\hat{\epsilon}_{cst}$ to obtain a crude estimate using OLS. After estimating Eq. 20 with the full set of fixed effects, I estimate $\hat{\rho}_{FE} = 0.53$, $P < 0.01$ ¹². A formal panel-test based on the residuals of the first-differenced model also rejects the null of no serial correlation in the error term $\hat{\rho}_{FD} = -0.011$, $P = 0.77$ and $P[\hat{\rho}_{FD} \neq -0.5] < 0.01$ ¹³. First-differencing in itself does not remove terms $\Delta\beta_{ct}$ and $\Delta\gamma_{st}$, corresponding to unobserved country-year or sector-year specific shocks from the equation, which may also be correlated with the explanatory variables and bias the coefficient estimates in the FD-equation. Running Hausman tests for the presence of these effects in the first-difference equation yields inconclusive results, the outcome depends on the method used to run the Hausman test. There is also a danger that putting fixed effects in a first-differenced equation estimated on data of not very high quality removes too much useful information and aggravates the impact of measurement error in the data, yielding attenuation bias.

The approach I will adopt following this discussion is to estimate 3 models: The simple FD specification, the FD specification with country-year and sector-year FE, and a FE specification with the full set of fixed effects.

These models are first estimated in log-level form as specified in Eq. 20, which gives the percentage change in VA in response to a 0.01 unit increase in the GVC indicators (I2E and E2R), expressed as a share of gross exports. But as Figure 35 shows, different sectors have vastly different I2E and E2R shares, thus for a sector with a low level of GVC integration, an increase in the share of 0.01 may imply a vastly greater degree of restructuring of production and potential gains or losses than a 0.01 increase for a sector already quite integrated in GVCs. To take account of these different levels of GVC integration, I run an additional set of 3 regressions where the log of the share instead of the share is included on the right-hand-side, thus giving an elasticity as the percentage change in VA from a percentage change in the share. Finally, since Kummritz (2016) uses the log of the values of I2E and E2R (not expressed as a percentage of exports but as VA content in exports), and taking the log of a large value may seem more natural than taking the log of a share, I also estimate 3 regressions with this classical elasticity specification. Thus in total I report 9 regressions: 3 different estimators times 3 different transformations of the independent variables. Table 9 reports the results¹⁴.

The coefficients from all specifications show a negative contemporaneous relationship between VA and I2E. This is probably a quite mechanical result by the nature of the close relation between VA and I2E being the foreign content in production and exports. A domestic shock of any form

¹⁰This test is based on an auxiliary specification $\tilde{y}_{it} = \tilde{X}_{it}\beta + \tilde{X}_{it}\gamma + \epsilon_{it}$ that can be estimated with robust standard errors, where \tilde{y}_{it} and $\tilde{X}_{it}\beta$ are the quasi-demeaned data for RE estimation and \tilde{X}_{it} are the time-demeaned predictors capturing the individual-variation in X . The test is an F-test of the exclusion restriction of \tilde{X}_{it} . If the test rejects, RE is likely inconsistent. See Wooldridge (2010) sec. 10.7.3.

¹¹ $\hat{\epsilon}_{cst}$ is not a clean estimate of ϵ_{cst} , but an estimate of the multiply-centered version of ϵ_{cst} .

¹²Wooldridge (2010) sec. 10.5.4 observes, under the null of no serial correlation in the errors, the residuals of a FE model must be negatively serially correlated, with $\text{cor}(\hat{u}_{it}, \hat{u}_{is}) = -1/(T-1) = -0.1$ with $T = 11$ in this case.

¹³This is the case because, for each $t > 1$, $\text{var}(\Delta u_{it}) = \text{var}(u_{it} - u_{i,t-1}) = \text{var}(u_{it}) + \text{var}(u_{i,t-1}) = 2\sigma^2$ with the assumptions of no serial correlation in u_t and constant variance. Because the residual has a zero mean and symmetric ACF, the covariance is $E[\Delta u_{it} \cdot \Delta u_{i,t+1}] = E[(u_{it} - u_{i,t-1})(u_{i,t+1} - u_{it})] = E[u_{it}u_{i,t+1}] - E[u_{it}^2] - E[u_{i,t-1}u_{i,t+1}] + E[u_{i,t-1}u_{it}] = -E[u_{it}^2] = -\sigma^2$, because of the no serial correlation assumption. Because the variance is constant across t , $\text{cor}(\Delta u_{it}, \Delta u_{i,t-1}) = \text{cov}(\Delta u_{it}, \Delta u_{i,t+1})/\text{var}(\Delta u_{it}) = -\sigma^2/2\sigma^2 = -0.5$.

¹⁴Estimation was done using the R package *fixest* (Bergé, 2018).

may cause VA to increase/decrease and the imported share to fall/rise in the current period. It is therefore more interesting to examine the lagged relationship of I2E and VA. Here the coefficients of the preferred FD specification signify a significant positive effect. In the first regression on the shares (S), only the second lag of I2E is significant at the 10% level, implying that a 0.01 unit increase in I2E is associated with a 0.31% increase in VA after two years. When using the log of the share (ES) however the coefficients on both lags are positive and significant, with a 1% increase in the I2E share associated with a 0.29% increase in VA after two years. The specification using the log of foreign VA in exports (E) yields that a 1% increase yields a 0.14% increase in VA after two years. The FD-TFE and FE specifications do not pick up an effect after one year but a larger effect after 2 years. As noted because of significant serial correlation FD is more efficient here¹⁵.

Table 9: VALUE ADDED REGRESSIONS

<i>Model:</i>	FD S (1)	FD-FE S (2)	FE S (3)	FD ES (4)	FD-FE ES (5)	FE ES (6)	FD E (7)	FD-FE E (8)	FE E (9)
(Intercept)	0.0716*** (0.0041)			0.0746*** (0.0029)			0.0390*** (0.0022)		
I2E	-1.593*** (0.3944)	-5.254*** (0.8604)	-5.354*** (0.7238)	-0.1323*** (0.0474)	-0.7345*** (0.1333)	-0.4249*** (0.1233)	-0.4931*** (0.0403)	-0.5564*** (0.0504)	-0.5420*** (0.0730)
L1.I2E	-0.0526 (0.3921)	-0.4503 (0.6753)	-0.8266 (1.379)	0.1756*** (0.0228)	0.1046 (0.0887)	0.0188 (0.1296)	0.1092*** (0.0223)	0.0208 (0.0353)	0.0092 (0.0488)
L2.I2E	0.3079* (0.1678)	0.6766 (0.5324)	1.307** (0.6186)	0.1114*** (0.0230)	0.0340 (0.0254)	-0.0127 (0.0521)	0.0265* (0.0143)	0.0436* (0.0229)	0.0630** (0.0241)
E2R	3.871*** (0.8564)	1.380*** (0.4795)	1.638** (0.6379)	0.9962*** (0.0388)	0.8492*** (0.0633)	0.8968*** (0.0596)	0.8271*** (0.0460)	0.8950*** (0.0242)	0.8585*** (0.0380)
L1.E2R	1.378*** (0.2895)	0.2976 (0.2985)	-0.2803 (0.4063)	0.1271*** (0.0464)	0.0349 (0.0323)	0.0862 (0.0756)	0.0800*** (0.0237)	-0.0073 (0.0266)	0.0487 (0.0567)
L2.E2R	0.4809** (0.1968)	-0.1346 (0.1963)	-0.7844** (0.3169)	0.0954*** (0.0262)	-0.0273 (0.0261)	-0.1421** (0.0606)	0.0244* (0.0136)	0.0039 (0.0197)	-0.0536 (0.0372)
<i>Fixed-Effects:</i>									
cs (N)	—	—	100	—	—	100	—	—	100
cy (N)	—	40	45	—	40	45	—	40	45
sy (N)	—	160	180	—	160	180	—	160	180
Cluster SE	cs	cs cy sy	cs cy sy	cs	cs cy sy	cs cy sy	cs	cs cy sy	cs cy sy
Observations	800	800	900	798	798	898	798	798	898
R ²	0.301	0.802	0.998	0.701	0.961	1.00	0.758	0.970	1.00
Within R ²		0.483	0.467		0.898	0.881		0.921	0.900

Note: The dependent variable is the natural log of VA, which is regressed on the export shares (1)-(3), natural log of the shares (4)-(6), and natural log of values (7)-(9) of I2E and E2R. *p<0.1; **p<0.05; ***p<0.01

For E2R, the results imply a large contemporaneous relationship with VA with an elasticity around 1. Also in this case caution needs to be exerted towards interpreting this as a structural shift in production, it could be for example that supply chain shocks contemporaneously lead all participating countries to export more and thus trigger an increase in both VA and E2R. However the contemporaneous relationship between VA and the E2R is less obvious than the relationship between VA and I2E. The lagged coefficients on E2R for the FD S equation also show a large impact, implying that that a 0.01 unit increase in E2R yields a 1.86% gain in GDP growth over two years. Curiously however the elasticity specifications show a much weaker impact, with the elasticity of the share (FD-ES) implying that a 1% increase in E2R gives a 0.22% growth gain within two years, and the classical elasticity (FD E) implying that a 1% increase in the foreign VA in exports yields a 0.1% increase in growth within two years. The FD-FE and FE specifications show mostly insignificant results on the lags, with the exception of two significant and negative effects for the FE S and FE ES specifications. These could also be the result of some attenuation bias; the quality of this data after removing all cross-country, sectoral and temporal variation between countries and sectors is likely not very high. I note that interpreting only the coefficients on the

¹⁵I have argued before for the likely consistency of all estimators on the basis of inconclusive Hausman Tests for the inclusion of fixed effects in the first difference equation.

lags as causal, *ceteris paribus* any contemporaneous causal effect, implies that the combined effect of the lags constitutes a lower bound estimate of the true effect of GVC integration on growth.

It should also be noted here that there are some significant outliers affecting especially the log-level regressions. The omission of these does not dramatically change the conclusion of the analysis but can shift the relative magnitude of coefficients on the lagged values a bit. For example excluding the 6 most influential data points in the first FD specification yields coefficients 0.25 and 0.43 on L1.I2E and L2.I2E.

To examine the effects of outliers on the coefficients in a more comprehensive manor, the regressions are re-estimated using a robust MM estimation method following [Yohai \(1987\)](#) and [Koller & Stahel \(2011\)](#), that downweights outliers and high-leverage data points using a highly efficient Iteratively Reweighted Least Squares (IRLS) procedure with 50% breakdown point and 95% asymptotic efficiency for normal errors¹⁶. The robust coefficient estimates are reported in Table 10. In preparation for estimation, the input data have been suitably differenced / demeaned to yield the same least squares estimates as reported in Table 9¹⁷.

Table 10: VALUE ADDED REGRESSIONS: ROBUST MM ESTIMATES

<i>Model:</i>	FD S (1)	FD-FE S (2)	FE S (3)	FD ES (4)	FD-FE ES (5)	FE ES (6)	FD E (7)	FD-FE E (8)	FE E (9)
(Intercept)	0.0722*** (0.0031)	-0.00002 (0.0011)	-0.0012 (0.0012)	0.0712*** (0.0026)	0.0006 (0.0007)	-0.0001 (0.0010)	0.0305*** (0.0022)	0.0006 (0.0006)	-0.0001 (0.0008)
I2E	-0.9218*** (0.2354)	-3.9477*** (0.3435)	-5.3803*** (0.2889)	-0.1070*** (0.0345)	-0.7019*** (0.0604)	-0.4799*** (0.0795)	-0.4988*** (0.0335)	-0.5316*** (0.0232)	-0.4942*** (0.0352)
L1.I2E	0.3766** (0.1608)	0.1692 (0.1661)	0.6597** (0.2827)	0.1457*** (0.0295)	0.0883* (0.0533)	-0.0906 (0.0827)	0.0688*** (0.0258)	0.0214 (0.0196)	-0.0325 (0.0351)
L2.I2E	0.4294*** (0.1381)	0.1286 (0.1772)	0.7084*** (0.2108)	0.1208*** (0.0241)	-0.0046 (0.0251)	0.0055 (0.0340)	0.0401*** (0.0124)	0.0309** (0.0145)	0.0420** (0.0187)
E2R	4.8046*** (0.4523)	0.8824*** (0.3124)	1.1079*** (0.2357)	1.0401*** (0.0344)	0.8666*** (0.0224)	0.9033*** (0.0295)	0.8661*** (0.0348)	0.9034*** (0.0098)	0.8794*** (0.0199)
L1.E2R	1.1457*** (0.2623)	0.2295 (0.1544)	-0.0044 (0.2139)	0.1195** (0.0544)	0.0047 (0.0162)	0.0270 (0.0371)	0.0673** (0.0320)	-0.0060 (0.0138)	0.0147 (0.0300)
L2.E2R	0.8208*** (0.2061)	-0.0220 (0.1328)	-0.4501*** (0.1403)	0.0885*** (0.0224)	-0.0223 (0.0177)	-0.0902** (0.0350)	0.0145 (0.0146)	0.0040 (0.0104)	-0.0231 (0.0169)
<i>Fixed-Effects:</i>									
cs (N)	–	–	100	–	–	100	–	–	100
cy (N)	–	40	45	–	40	45	–	40	45
sy (N)	–	160	180	–	160	180	–	160	180
SE	HAC	HAC	HAC	HAC	HAC	HAC	HAC	HAC	HAC
Observations	800	800	900	798	798	898	798	798	898
R ²	0.392	0.588	0.729	0.763	0.924	0.899	0.858	0.951	0.924
Adjusted R ²	0.387	0.584	0.727	0.762	0.923	0.898	0.857	0.950	0.924
Residual SE	0.068	0.026	0.033	0.055	0.019	0.026	0.040	0.016	0.023
IRLS Covered	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes

Note: The dependent variable is the natural log of VA, which is regressed on the export shares (1)-(3), natural log of the shares (4)-(6), and natural log of values (7)-(9) of I2E and E2R. *p<0.1; **p<0.05; ***p<0.01

In comparison to the least squares estimates in Table 9, the coefficients in Table 10 spread the effect more evenly across the lags in the FD specification, and also let the FD-FE and FE specifications move closer to the plain FD specification. Notably, the coefficient in L1.I2E in the log-level specification is now positive and significant. The results for the FD S and FD ES specifications imply that a 0.01 / 1% increase in I2E results in a 0.8% / 0.27% increase in VA after two years. The coefficients on the FD E specification are slightly smaller than in Table 9, with a 1% increase in the value of I2E resulting in an 0.11% increase in VA within two years. Similar

¹⁶Estimation is done by a robust MM procedure using IRLS with a bi-square redescending score function, resulting in a highly robust and highly efficient estimator (with 50% breakdown point and 95% asymptotic efficiency for normal errors), Implemented in the R package *robustbase* ([Rousseeuw et al., 2020](#)).

¹⁷Multi-dimensional centering and differencing in preparation for robust estimation was done using the R package *collapse* ([Krantz, 2021](#)).

distributional effects hold for E2R, where additionally the combined semi-elasticity from FD S is around 2 and thus a bit larger than the 1.86 in Table 9, while the elasticities are a bit smaller with 0.21 / 0.082 from FD ES / FD E compared to 0.22 / 0.1 in Table 9. In terms of robustness, it should also be noted here that multicollinearity between the various lagged values is low, at a maximum VIF of 1.3 in all models.

5.1.1 Manufacturing Estimates

Tables 11 and 12 reports equivalent regressions run for the manufacturing sub-sample of sectors. Compared to Tables 9 and 10 the results appear to be broadly similar and coefficients slightly larger.

Table 11: VALUE ADDED REGRESSIONS: MANUFACTURING

Model:	FD S (1)	FD-FE S (2)	FE S (3)	FD ES (4)	FD-FE ES (5)	FE ES (6)	FD E (7)	FD-FE E (8)	FE E (9)
(Intercept)	0.0678*** (0.0088)			0.0752*** (0.0038)			0.0419*** (0.0041)		
I2E	-1.488** (0.5558)	-5.526*** (1.892)	-5.952** (2.449)	-0.1546* (0.0901)	-0.8395*** (0.2898)	-0.6364** (0.2812)	-0.5477*** (0.0491)	-0.4025*** (0.0599)	-0.3972*** (0.0954)
L1.I2E	-0.9133 (0.5808)	-3.216 (3.368)	-5.472 (5.684)	0.1360*** (0.0337)	-0.3627** (0.1647)	-0.6820*** (0.1900)	0.0942* (0.0466)	-0.0548 (0.0388)	-0.0986* (0.0567)
L2.I2E	0.5380** (0.2403)	2.480 (1.625)	6.195 (4.257)	0.1782*** (0.0506)	0.3872 (0.2602)	0.4091* (0.2211)	0.0590** (0.0252)	0.0175 (0.0249)	0.1184** (0.0568)
E2R	3.990** (1.479)	0.4527 (0.4625)	1.434 (1.001)	0.9845*** (0.0505)	0.8140*** (0.0867)	0.8362*** (0.0643)	0.8882*** (0.0469)	0.9013*** (0.0456)	0.8926*** (0.0478)
L1.E2R	2.446*** (0.4453)	-0.2512 (0.4383)	-0.8829 (0.9402)	0.2056** (0.0771)	0.0725 (0.0519)	0.1013 (0.0837)	0.1228** (0.0455)	0.0184 (0.0230)	0.0883 (0.0838)
L2.E2R	0.2386 (0.2851)	-0.3984 (0.3696)	-0.8486 (0.5726)	0.0417 (0.0337)	-0.0192 (0.0230)	-0.1399 (0.0908)	-0.0097 (0.0197)	0.0252 (0.0281)	-0.1072 (0.0894)
<i>Fixed-Effects:</i>									
cs (N)	—	—	40	—	—	40	—	—	40
cy (N)	—	40	45	—	40	45	—	40	45
sy (N)	—	64	72	—	64	72	—	64	72
Cluster SE	cs	cs cy sy	cs cy sy	cs	cs cy sy	cs cy sy	cs	cs cy sy	cs cy sy
Observations	320	320	360	320	320	360	320	320	360
R ²	0.309	0.837	0.995	0.813	0.979	0.999	0.849	0.987	1.00
Within R ²		0.173	0.178		0.892	0.925		0.936	0.940

Note: The dependent variable is the natural log of VA, which is regressed on the export shares (1)-(3), natural log of the shares (4)-(6), and natural log of values (7)-(9) of I2E and E2R. *p<0.1; **p<0.05; ***p<0.01

Again especially the log-level specification is affected by outliers, thus under general considerations the robust estimates in Table 12 are preferable and also distribute the effect more evenly across lagged coefficients. In both Tables 11 and 12, the lagged coefficients on the FD-FE and FE specifications appear to be attenuated and are largely insignificant. Gauging thus from the simple first-difference lagged coefficients in Table 12, the semi-elasticity of manufacturing VA to a 0.01 unit increase in I2E / E2R is a 0.58% / 2.47% increase, the elasticity w.r.t. a 1% increase in I2E / E2R is 0.28% / 0.31%, and the elasticity w.r.t. a 1% increase in the values of I2E / E2R is 0.15% / 0.07%. Again these are lower bound effect sizes, *ceteris paribus* any contemporaneous effects.

5.1.2 Comparison of Estimates

In summary, Tables 11 and 12 present similar results than Tables 9 and 10. Using only the robust estimates from Tables 10 and 12 of the preferred FD specification yields that:

- A 0.01 unit increase in I2E / E2R yields a 0.81% / 1.97% increase in overall VA and a 0.58% / 2.47% increase in manufacturing VA after 2 years.

- A 1% increase in I2E / E2R yields a 0.27% / 0.21% increase in overall VA and a 0.28% / 0.31% increase in manufacturing VA after 2 years.
- A 1% increase in the values of I2E / E2R yields a 0.11% / 0.082% increase in overall VA and a 0.15% / 0.07% increase in manufacturing VA after 2 years.

Given that, as evident from Figures 34 and 35, manufacturing sectors have a higher I2E and lower E2R ratio than other sectors, and, as shown in Table 8 overall I2E in the sample with an average of 0.2 is higher than E2R with an average of 0.15, it appears natural that the semi-elasticity of VA w.r.t. E2R is higher than w.r.t. I2E, and that manufacturing sectors on average gain more from improvements in E2R.

Table 12: VALUE ADDED REGRESSIONS: MANUFACTURING: ROBUST MM ESTIMATES

<i>Model:</i>	FD S (1)	FD-FE S (2)	FE S (3)	FD ES (4)	FD-FE ES (5)	FE ES (6)	FD E (7)	FD-FE E (8)	FE E (9)
(Intercept)	0.0718*** (0.0045)	-0.0038** (0.0015)	-0.0040 (0.0024)	0.0728*** (0.0038)	0.0006 (0.0010)	-0.0002 (0.0013)	0.0310*** (0.0034)	0.0007 (0.0008)	-0.0004 (0.0012)
I2E	-0.5597 (0.3580)	-2.4579*** (0.5795)	-5.9545*** (1.1946)	-0.0545 (0.0678)	-0.8506*** (0.1373)	-0.7222*** (0.1886)	-0.5498*** (0.0430)	-0.4258*** (0.0408)	-0.4078*** (0.0832)
L1.I2E	0.1038 (0.2566)	0.3573 (0.6005)	0.8177 (1.1756)	0.1185*** (0.0455)	-0.2473** (0.1252)	-0.4732* (0.2637)	0.0835** (0.0377)	-0.0626* (0.0362)	-0.0972 (0.0654)
L2.I2E	0.4707** (0.2376)	0.8345** (0.4161)	1.4843** (0.7338)	0.1592*** (0.0458)	0.2528* (0.1407)	0.1105 (0.1830)	0.0639*** (0.0195)	-0.0153 (0.0166)	0.0581* (0.0320)
E2R	6.2357*** (0.5858)	0.1768 (0.1489)	0.8435** (0.3522)	0.9901*** (0.0465)	0.8083*** (0.0460)	0.8488*** (0.0543)	0.9322*** (0.0371)	0.8903*** (0.0235)	0.9058*** (0.0331)
L1.E2R	1.6517*** (0.3458)	-0.0723 (0.1590)	-0.2776 (0.4348)	0.2336*** (0.0441)	0.0246 (0.0215)	0.0897 (0.0712)	0.0690* (0.0409)	0.0109 (0.0136)	0.0729** (0.0336)
L2.E2R	0.8143*** (0.2503)	-0.1095 (0.1171)	-0.5303* (0.3107)	0.0731*** (0.0210)	-0.0168 (0.0350)	-0.0802 (0.0919)	-0.0052 (0.0199)	0.0166 (0.0121)	-0.0416 (0.0259)
<i>Fixed-Effects:</i>									
cs (N)	—	—	40	—	—	40	—	—	40
cy (N)	—	40	45	—	40	45	—	40	45
sy (N)	—	64	72	—	64	72	—	64	72
SE	HAC	HAC	HAC	HAC	HAC	HAC	HAC	HAC	HAC
Observations	320	320	360	320	320	360	320	320	360
R ²	0.539	0.136	0.299	0.871	0.907	0.956	0.936	0.944	0.961
Adjusted R ²	0.530	0.119	0.287	0.869	0.905	0.955	0.934	0.942	0.960
Residual SE	0.066	0.024	0.039	0.049	0.016	0.021	0.038	0.013	0.020
IRLS Covered	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes

Note: The dependent variable is the natural log of VA, which is regressed on the export shares (1)-(3), natural log of the shares (4)-(6), and natural log of values (7)-(9) of I2E and E2R. *p<0.1; **p<0.05; ***p<0.01

Larger productivity gains from forward integration proxied by E2R is also a salient feature in the literature. Kummritz (2016), using a sample of mostly manufacturing industries, finds robust benefits of GVC backward and forward integration on VA in both developing and developed / middle income countries, with a larger benefit of forward integration (E2R) at elasticities as high as 0.58 for developing / middle income countries and 0.68 for developed countries, and I2E elasticities smaller around 0.2-0.3. He also estimates labour productivity elasticities to E2R of 0.29 for developing / middle-income countries and 0.49 for developed countries.

In a similar exercise Kummritz (2015) finds that high-income countries benefit relatively more from forward linkages (E2R) whereas middle-income countries also benefit from backward linkages (I2E). The results presented here suggest that EAC countries could benefit almost equally from increases in backward and forward GVC integration, with manufacturing sectors drawing slightly greater benefits from forward integration. This is a sensible result, as for example increased forward integration in primary products like agriculture or mining is not necessarily associated with domestic productivity gains, but if manufactured exports are re-exported as part of a value chain, they likely have to be of sufficient quality.

6 Summary

This study set out to undertake a broad and comprehensive analysis of the patterns of global and regional value chain integration in the EAC, using global MRIO tables provided by EORA - the only ICIO database covering the EAC at the time of writing. The decade 2005-2015 were chosen as analysis period, mandated by the full EAC membership of Rwanda and Burundi in 2007, and the availability of data only up to 2015.

The paper commenced with a broad exposition of gross input-output (IO), exports and value added (VA) flows within the EAC, and between the EAC and different World regions. This exposition was followed by an econometric analysis seeking to quantify the contribution of GVC integration to member countries GDP growth and industrialization. The results and discussion are summarized below.

6.1 Gross Flows

The general structure of production in most EAC countries is that domestic VA (GDP) constitutes about 50-60% of gross output. The remainder of gross output are intermediate inputs, of which in most EAC countries 5-10% are imported. In all EAC countries apart from Kenya, 5% or less of gross output are exported by 2015, in Kenya about 8%. In Uganda about 25% of imported inputs are from the EAC, followed by Rwanda at 14%, and the other countries below 10% (Kenya around 2%). In 2015 about 30% of Ugandan and Kenyan exports are to the EAC, whereas the EAC export shares of all other members are below 7%.

The analysis of gross IO flows highlighted the supplier role of Kenya within the EAC, followed with some distance by Uganda and Tanzania, and a negligible role of Rwanda and Burundi. The largest external supplier of inputs to all EAC countries is the European Union, followed, again with some distance, by South Asia (SAS), the Middle East & North Africa (MEA), the rest of Sub-Saharan Africa (SSA) and China.

The EAC trade balance in gross IO flows is negative, and has deteriorated over time: In 2005 intermediate imports were 2x larger than intermediate exports, by 2015 they were 3x larger.

At the sector-level, the largest outgoing flows are agricultural inputs, especially from Kenya, but also from Tanzania and Uganda, into EU food processing and beverage industries as well as agricultural re-exports by the EU. Among the incoming flows, the EU, and also MEA, SAS and SSA supply inputs for EAC transport, construction, petro-chemical and electrical machinery sectors. Inside the EAC, Kenya supplies mining and manufacturing inputs to Ugandan and Tanzanian manufacturing (esp. petro-chemicals and electrical machinery), transport and construction sectors, and Uganda supplies agricultural inputs to the Kenyan food processing industry. At a significantly lower scale Uganda also supplies manufacturing inputs for Rwandan manufacturing (esp. petro-chemicals, metal products and electrical machinery). Tanzania has even much less of a supplier flow, the only flow close to 1 million dollars in 2015 is Tanzanian mining inputs for the Ugandan petro-chemical industry. Rwanda and Burundi appear to be negligible as suppliers of inputs in EAC production, and South Sudan appears not to be economically integrated with the EAC on the production side at all.

Over the 2005-2015 period, ROW inputs into EAC production are quite steadily 12 times greater than EAC inputs into EAC production, whereas intermediate outflows to the ROW are only 4.4 times greater than inner-EAC flows in 2015, down from 5.9 times greater in 2005. This suggests an asymmetric development where the EAC is integrating but only in terms of intermediate outflows.

In terms of exports, in all major EAC countries more than 50% of exports is made up from agriculture, fishing, mining, and food and beverage industries. Uganda has the largest share of raw agricultural exports at 38% in 2015, Kenya and Tanzania have lower shares at 29% and 24%, respectively. In Kenya the food and beverages industry is quite important, comprising 14.3% of exports in 2015. Rwanda has a large share of 24% in mining exports. Apart from primary product,

Kenya also exports petro-chemicals (10.5%) and electrical machinery (6%). Tanzania also exports textiles and other manufactures (both at shares of 9%). For all EAC countries exported transport services also have a share between 5 and 10%.

Of the EAC members, only Uganda and Kenya export around 30% of their exports to the EAC. Ugandan exports to the EAC are comprised to more than 60% of primary agriculture, and more than 50% of Kenyan exports to the EAC are manufactured products such as petro-chemical, metal products and electrical machinery. Tanzania exports 7% to the EAC, more than 20% of which is foods and beverages. Rwanda, Burundi and South Sudan export less than 2% to the EAC.

6.2 Value Added Flows

In VA terms, deriving the value-added-multiplier matrix **VB** at both aggregate and sectoral levels yielded that in the aggregate, 85-90% of the value of aggregate output is sourced within the country for most EAC countries. Notable exceptions were Tanzania at 68% of domestic VA and South Sudan at 98% domestic VA (in 2015).

As largest foreign supplier of VA in 2015, the EU adds 2.38% to Uganda's, 7.49% to Tanzania's, 4.58% to Kenya's 3.77% to Rwanda's, 2.77% to Burundi's and 0.11% to South Sudan's production. Other significant suppliers of VA in EAC production are South Asia, Sub-Saharan Africa, the Middle East and China. Among the EAC countries, Kenya takes on a significant role, supplying 2.24% in Uganda's, 1.7% in Tanzania's and 0.83% in Rwanda's production. Uganda also adds 0.38% in Rwanda's production.

The foreign VA share in a countries final goods production (and exports) was termed vertical specialization (VS) by [Hummels et al. \(2001\)](#), and I2E by [Baldwin & Lopez-Gonzalez \(2015\)](#), and is a widely used measure of backwards integration into GVCs. Computing this measure for the years 2005-2015 yielded little evidence that EAC countries have increased their participation in GVCs. In Uganda, VS has been fluctuating between 10 and 12% without a clear trend. In Tanzania VS has increased remarkably from 13% in 2005 to 29% in 2015, but this result should be treated with extreme caution as the macro-data for Tanzania is highly distorted¹⁸. In Kenya VS has been oscillating around 15%, in Rwanda around 12%, in Burundi around 10%. A notable feature of the data is that VS appears to have increased in Uganda, Kenya, Rwanda and Burundi in the years 2005-2011, and decreased again from 2012-2015.

In most EAC countries EAC VA shares (in particular VA by Kenya and Uganda) have increased by small amounts of 0-0.5 percentage points over the analyzed period, indicating that the relative importance of EAC neighbours as suppliers of inputs has increased.

At the sector level, VS is highest in manufacturing sectors in all EAC countries, coming as high as 60% in some Tanzanian sectors (which should be regarded with caution). In Uganda, Kenya, Rwanda and Burundi, between 20% and 40% of VA in core manufacturing industries such as petro-chemicals, electrical machinery, metal products and transport equipment is foreign sourced. Recycling and re-export industries in a number of EAC countries also have high VS of 70% and above. Again most of the foreign VA in EAC production comes from the EU, which also quite homogeneously applies at the sector level.

The **VB** matrix was also used to decompose IO flows into VA terms, yielding a more favorable but still very negative EAC trade balance in intermediates. This is a result of IO flows from ROW to the EAC manufacturing industries having less domestic VA, than EAC intermediate exports to ROW comprising mostly of primary inputs. Also in VA terms intermediate flows inside the EAC are around 12 times smaller than intermediate inflows from ROW to EAC. Progress towards greater regional integration in intermediates has only been made on the outflow side, with the relative importance of ROW declining from 7.4 times greater in 2005 to 5.5 times greater in 2015.

¹⁸Tanzanian GDP for example is decreasing as shown in Figure 18, which calls into question the reliability of IO relationships and shares computed for Tanzania.

VA exports (comprising of both intermediate outflows and final goods) increased in all EAC countries between 2005 and 2015, with Ugandan VA exports estimated at 1 billion USD at basic prices in 2015, and Kenyan VA exports around 6 billion.

A popular measure of forward GVC integration is the VA share of gross exports that enters foreign countries exports, termed VS1 by [Hummels et al. \(2001\)](#) and E2R by [Baldwin & Lopez-Gonzalez \(2015\)](#). In most EAC countries VS1 has been quite stable over time, in Uganda it has dropped from 14% in 2005 to 12% in 2015, in Kenya it has been stable at 12.5%, in Rwanda it dropped from 23% to 16%, in Burundi it is stable around 17%. Thus overall the EAC GVC situation appears quite stable, no country has progressed to supply significant value for foreign export productions.

At the sector level, mostly primary inputs and wholesale traded goods are re-exported by receiving countries. In Uganda 48% of mining exports are re-exported, and 30% of wholesale goods exports. In Kenya, Rwanda and Burundi, apart from wholesale trade, 20% and more of agricultural exports are being re-exported. At the sector level, some development is visible in the 2005-2015 period: most manufacturing sectors increased their imported content (VS) and decreased their re-exported content (VS1), whereas in most primary sectors (mainly agriculture), the opposite is the case. Overall this suggests that EAC countries have failed to move upstream into manufacturing GVCs, and are increasingly concentrating on agriculture and the production of simple manufactures for domestic consumption and final export.

On the regional front, computing the EAC partner contribution to each members VS, VS1, value-added imports (VAI) and imports of final goods (VAFI) yielded moderate declines in most cases. The only marked increase was the EAC share in Kenya's VS1, which increased from 5.7% in 2005 to 7.4% in 2015, highlighting Kenya's increasing supplier role for EAC export production. It should be noted that these shares are at very different levels for different EAC countries. In Uganda, the EAC accounts for 21% of VS, 6% of VS1 and 17% of VAI. Thus Uganda primarily relies on the EAC for inputs, whereas Kenya, with an EAC share in VS of only 0.9%, primarily exports inputs to EAC countries.

Thus in overall terms, with the exception on Tanzania and Rwanda which seem to have become globally more integrated as users of inputs for export production (VS), and Kenya who has increased its role as an EAC supplier of inputs (EAC share in VS1), it appears that the decade under analysis has seen little progress towards integration into global or regional value chains.

Advanced analysis of GVC integration using the [Koopman et al. \(2014\)](#) (short KWW) decomposition of gross exports yielded that there is hardly any double counted domestic VA in gross exports, which arises from two-way trade in intermediate goods and constitutes a meaningful fraction of exports in advanced economies. This implies that EAC countries only engage in uni-directional (and mostly shorter) GVCs. Nevertheless up to 10% of VA exported is foreign double counted, implying the re-importing and exporting of foreign VA is prevalent. There is also practically 0 domestic VA that returns via final or intermediate imports, implying again shallow GVC integration and relatively uni-directional trading relationships.

Indices of upstreamness as the share of domestic VA in intermediate exports over total DVA and downstreamness as foreign VA in final exports over total FVA, showed that, apart from Tanzania where the data suggest a slight upstream movement, all EAC countries have moved downstream in GVCs, with both less domestic content going into intermediate exports and more FVA going to final goods exports.

Using KKW to decompose EAC exports to the EAC yields more domestic and foreign VA in final goods compared to EAC exports to ROW. Also double counted and re-exported components are lower, confirming that EAC countries engage more in GVCs with ROW than with their EAC neighbours.

Using the composition developed by [Wang et al. \(2013\)](#), exports were also decomposed at the sector level. At the sector level it is again evident that manufacturing sectors have higher VS

shares than other sectors, and also higher amounts of double counting, although mostly of foreign VA. Calculations of upstreamness and downstreamness ratios at the sector level reveal that nearly all sectors in all EAC countries have moved downstream between 2005 and 2015. This is surprising, indicating that in terms of GVC integration there are little exceptions or high-performing sectors, and the downstream trends observed in the aggregate dissipate quite uniformly to the sector level.

A final exercise in the exploration of VA flows was to compute the New Revealed Comparative Advantage (NRCA) index, using GDP in exports as proposed by [Koopman et al. \(2014\)](#). This revealed a comparative advantage in agriculture and fishing, and a comparative disadvantage in manufacturing for all EAC members. On the services side, in all members apart from Kenya, construction, maintenance and repair activities, accommodation and food services (including Kenya) and activities of private households (self-employment) also have quite a strong NRCA. Kenya has a notable comparative advantage in food and beverages.

In the 2005-2015 period, NRCA in agriculture has remained stable, whereas all members lost NRCA in mining. All members also have gained strongly in re-exporting goods. Other country and sector developments are more heterogeneous. Uganda and Tanzania have gained NRCA in fishing, Kenya notably has gained NRCA in all manufacturing sectors.

To better expose potential issues and challenges of regional integration, NRCA has also been computed for EAC members relative to the combined EAC export mix. The core result of this was that relative to other EAC members Uganda has a NRCA in agriculture, Tanzania in fishing, Rwanda in mining, and Kenya in core manufacturing sectors such as wood and paper, petrochemicals, metal products and electrical machinery as well as food and beverages. Furthermore Rwanda and Burundi, and to a weaker extent Uganda, have a NRCA in construction, maintenance and repairs, wholesale and retail trade, whereas Tanzania appears to have a NRCA in other manufacturing, recycling, and financial and business services.

6.3 GVCs and Industrial Development

The exposition of gross and particularly of VA flows suggests that GVC integration has been sluggish, and even declined in most EAC countries. The nature of integration has also shifted towards more downstream production resulting in the export of final goods using domestic and foreign VA. Since previous research has found GVCs to be beneficial for growth, particularly in advanced and middle income countries, and since most transfer of technology is regarded as taking place in the upstream parts of the value chains (rather than assembly tasks or supplying primary inputs), it remains to investigate whether increasing GVC integration could benefit the EAC.

Towards this end an econometric analysis was conducted regressing VA (GDP) on I2E (VS) and E2R (VS1) at the sector level. This analysis rested on the assumptions that sector-level heterogeneity can provide some information about the likely benefit of GVCs on aggregate economic activity, and that changes in I2E and E2R benefit economic activity with a lag of up to two years (which is the main identification assumptions apart from the use of first differences and fixed effects to eliminate cross-sectional heterogeneity). Robust statistical methods were used to minimize the effects of outliers in individual sectors on the result. Various effect sizes were obtained by regressing the log of VA on the export shares, the log of the shares, and the log of the values of I2E and E2R. Furthermore, additional regressions were run only for a subsample of core manufacturing sectors to obtain an estimate of the productivity impact of GVC integration on manufacturing.

The coefficients from all specifications show a strong negative contemporaneous relationship between VA and I2E and a strong positive relationship between VA and E2R. These contemporaneous relationships were however regarded as quite mechanical and were disregarded in favor of the coefficients on the lags of I2E and E2R.

The focus for interpretation was on the lagged relationships, which give a lower bound effect size estimate conditional on the contemporaneous relationship, but can carefully be interpreted as causal. Results from robust estimation of the preferred first-difference specification¹⁹ show a

¹⁹First-differences was preferred over fixed effects due to significant serial correlation in the error term.

moderate positive effect of both I2E and E2R on VA. About 70% of this effect materializes after once year while the remaining approx. 30% materializes in the second year following the increase. estimation of the elasticity of VA w.r.t. the I2E and E2R shares yielded at a 1% increase in I2E yields a 0.27% increase in VA after two years and a 1% increase in E2R yields a 0.21% increase in VA after two years. For manufacturing sectors the responses to I2E is 0.28% and to E2R is 0.31%. The overall larger impact on manufacturing, and particularly the importance of forward GVC integration as measured by E2R, emphasizes the potential of GVCs to boost productivity in EAC manufacturing sectors.

Larger gains from forward integration were also estimated in the literature such as [Kummritz \(2016\)](#) with manufacturing-heavy samples of sectors in OECD ICIO tables. The elasticities on forward GVC integration in these high-income countries is higher at around 0.6, whereas backup GVC integration (I2E) has a significantly lower elasticity (0.1-0.3) in high-income countries. Another salient finding by [Kummritz \(2015\)](#) is that low- and middle-income countries generally benefit less from GVC integration, and benefit relatively more from backward linkages (I2E) compared to high-income countries. These findings appear to be confirmed by the empirical results of this paper.

7 Conclusion

In spite of severe data limitations for EAC countries, in particular for Tanzania, by virtue of using a global MRIO model where data inconsistencies in small countries data are scaled away and time series interpolation is used to fill gaps in the data, this analysis has produced a few viable and important findings that are broadly consumerable with the observable EAC production and trading patterns, and with GVC analysis conducted for other low-and middle-income countries covered by higher quality ICIO tables such as OECD-UNCTAD and WIOD.

The most important of these findings are that in the years 2005-2015 EAC members do not seem to have integrated further into GVCs, both in overall terms and in terms of production sharing within the EAC. Moderate overall developments are visible in Rwanda and Tanzania which have gradually increased the foreign content in their production (VS), and on the regional level Kenya has become an important supplier of inputs which also feed into export production of most EAC members (E2R). Yet inputs to EAC production from the rest of the World remain consistency 12-times greater in VA terms than inputs from EAC neighbours.

Furthermore there appears to be a downstream shift in existing GVC relationships, with more domestic and foreign VA going into final goods production for domestic use and final export, while maintaining high levels of primary input exports such as agriculture. This movement comes at the cost of producing high-quality intermediate inputs which would allow EAC countries to integrate in upstream parts of GVCs where most efficiency gains and use of more complex technology occur. This downstream shift appears to be prevalent across EAC sectors, including manufacturing.

Econometric analysis provided evidence that increased GVC integration can benefit growth in the EAC. The elasticities of domestic VA to the two main GVC ratios (I2E and E2R) are around 0.3, which is a lower-bound estimate derived from the 1 and 2-year lagged coefficients in reduced from panel-data models. This impact estimate is slightly higher for manufacturing, which would also benefit slightly more from forward integration (E2R), in-line with the global evidence (e.g. [Kummritz \(2016\)](#), [Kummritz \(2015\)](#), [Foster-McGregor et al. \(2015\)](#)).

These findings suggest that policies that would help EAC industries to integrate more into regional and global value chains, through the production of high-quality intermediate inputs, would benefit EAC growth and productivity in the medium to long term.

It remains at this point to conduct further research to confirm and investigate the nature and causes of the downstream shift towards final goods in manufacturing observed by this study. It also remains to research further the more general limitations inhibiting EAC industries to engage in GVCs and RVCs, relating e.g. to the business environment and trade policies.

Finally, a lot more research can be done into the effects of GVCs on productive restructuring in the EAC and Sub-Saharan African economies more generally. This includes effects on product and export diversification and competitiveness, labour productivity and employment, and the specific details of where and how much value is added in different sectors as they become permeated by GVCs.

References

- Balassa, B. (1965). Trade liberalisation and “revealed” comparative advantage 1. *The manchester school*, 33(2), 99–123.
- Baldwin, R., & Lopez-Gonzalez, J. (2015). Supply-chain trade: A portrait of global patterns and several testable hypotheses. *The World Economy*, 38(11), 1682–1721.
- Baldwin, R., & Robert-Nicoud, F. (2014). Trade-in-goods and trade-in-tasks: An integrating framework. *Journal of international Economics*, 92(1), 51–62.
- Benz, S., Larch, M., & Zimmer, M. (2015). Trade in ideas: Outsourcing and knowledge spillovers. *International Economics and Economic Policy*, 12(2), 221–237.
- Bergé, L. (2018). Efficient estimation of maximum likelihood models with multiple fixed-effects: the R package FENmlm. *CREA Discussion Papers*(13).
- Beverelli, C., Stolzenburg, V., Koopman, R. B., & Neumueller, S. (2019). Domestic value chains as stepping stones to global value chain integration. *The World Economy*, 42(5), 1467–1494.
- Daly, J., Hamrick, D., Gereffi, G., & Guinn, A. (2016). Maize value chains in east africa. *Duke Center on Globalization, Governance, and Competitiveness*.
- Daudin, G., Riffart, C., & Schweisguth, D. (2011). Who produces for whom in the world economy? *Canadian Journal of Economics/Revue canadienne d’économique*, 44(4), 1403–1437.
- Foster-McGregor, N., Kaulich, F., & Stehrer, R. (2015). Global value chains in africa.
- Hausman, J. A. (1978). Specification tests in econometrics. *Econometrica: Journal of the econometric society*, 1251–1271.
- Hummels, D., Ishii, J., & Yi, K.-M. (2001). The nature and growth of vertical specialization in world trade. *Journal of international Economics*, 54(1), 75–96.
- Koller, M., & Stahel, W. A. (2011). Sharpening wald-type inference in robust regression for small samples. *Computational Statistics & Data Analysis*, 55(8), 2504–2515.
- Koopman, R., Wang, Z., & Wei, S.-J. (2014). Tracing value-added and double counting in gross exports. *American Economic Review*, 104(2), 459–94.
- Kowalski, P., Gonzalez, J. L., Ragoussis, A., & Ugarte, C. (2015). Participation of developing countries in global value chains: Implications for trade and trade-related policies.
- Krantz, S. (2021). collapse: Advanced and fast data transformation in R [Computer software manual]. Retrieved from <https://CRAN.R-project.org/package=collapse> (R package version 1.6.4)
- Kummritz, V. (2015). *Global value chains: benefiting the domestic economy?* (Tech. Rep.). Graduate Institute of International and Development Studies Working Paper.
- Kummritz, V. (2016). Do Global Value Chains Cause Industrial Development? *CTEI Working Paper No 2016-01*. Retrieved from <http://repec.graduateinstitute.ch/pdfs/cteiwp/CTEI-2016-01.pdf> doi: 10.1002/pd
- Kummritz, V., & Quast, B. (2014). decompr: global value chain decomposition in R.
- Kummritz, V., & Quast, B. (2016). Global value chains in low and middle income countries. *CTEI Working Papers*(10). Retrieved from <https://ideas.repec.org/p/gii/cteiwp/ctei-2016-10.html>
- Lenzen, M., Kanemoto, K., Moran, D., & Geschke, A. (2012). Mapping the structure of the world economy. *Environmental science & technology*, 46(15), 8374–8381.
- Lenzen, M., Moran, D., Kanemoto, K., & Geschke, A. (2013). Building eora: a global multi-region input-output database at high country and sector resolution. *Economic Systems Research*, 25(1), 20–49.

- Leontief, W. W. (1936). Quantitative input and output relations in the economic systems of the united states. *The review of economic statistics*, 105–125.
- Li, B., & Liu, Y. (2015). *Moving up the value chain* (Tech. Rep.). Mimeo, Boston University.
- Piermartini, R., & Rubínová, S. (2014). *Knowledge spillovers through international supply chains* (Tech. Rep.). The Graduate Institute of International and Development Studies.
- Rousseeuw, P., Croux, C., Todorov, V., Ruckstuhl, A., Salibian-Barrera, M., Verbeke, T., ... Maechler, M. (2020). Robustbase: basic robust statistics. *R package version 0.93-6*, URL <http://CRAN.R-project.org/package=robustbase>.
- Timmer, M., Erumban, A. A., Gouma, R., Los, B., Temurshoev, U., de Vries, G. J., ... others (2012). *The world input-output database (wiod): contents, sources and methods* (Tech. Rep.). Institutue for International and Development Economics.
- Wang, Z., Wei, S.-J., & Zhu, K. (2013). *Quantifying international production sharing at the bilateral and sector levels* (Vol. No. w19677; Tech. Rep.). National Bureau of Economic Research.
- Wooldridge, J. M. (2010). *Econometric analysis of cross section and panel data*. MIT press.
- Yohai, V. J. (1987). High breakdown-point and high efficiency robust estimates for regression. *The Annals of Statistics*, 642–656.