

Tackling supply chain 'blind spots' via end-to-end visibility, traceability, and transparency: Conceptualising the role publicly data and big data analytics

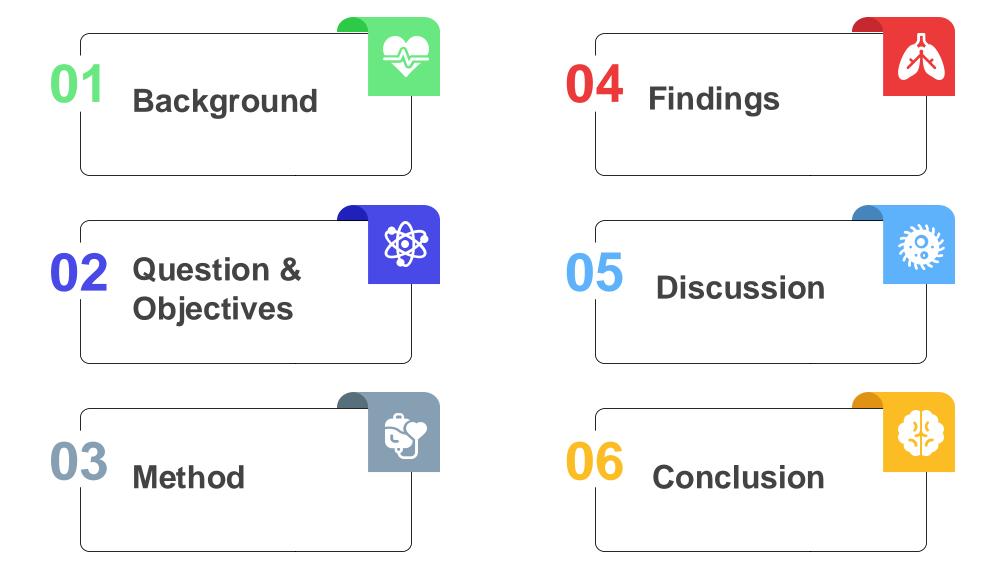
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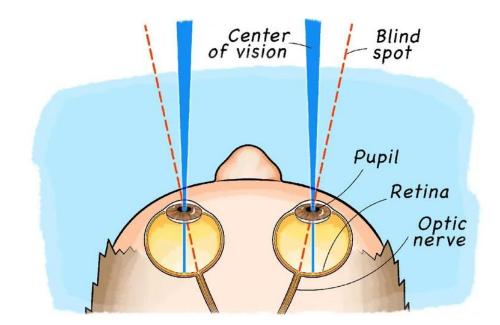
Agenda

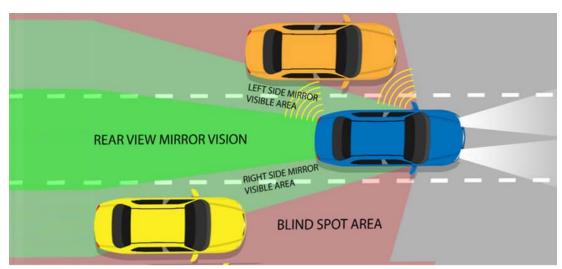


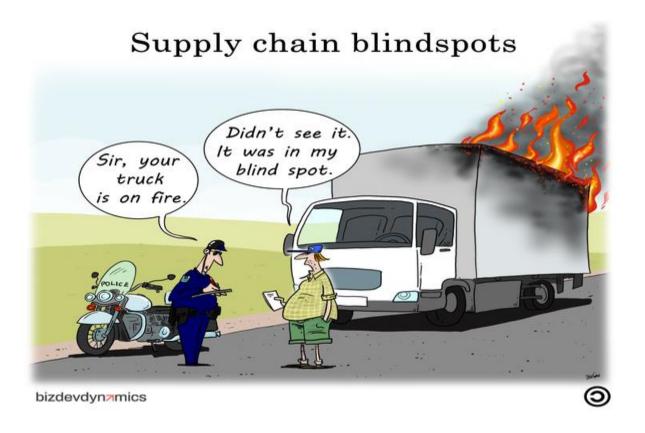




Background – Blind Spots







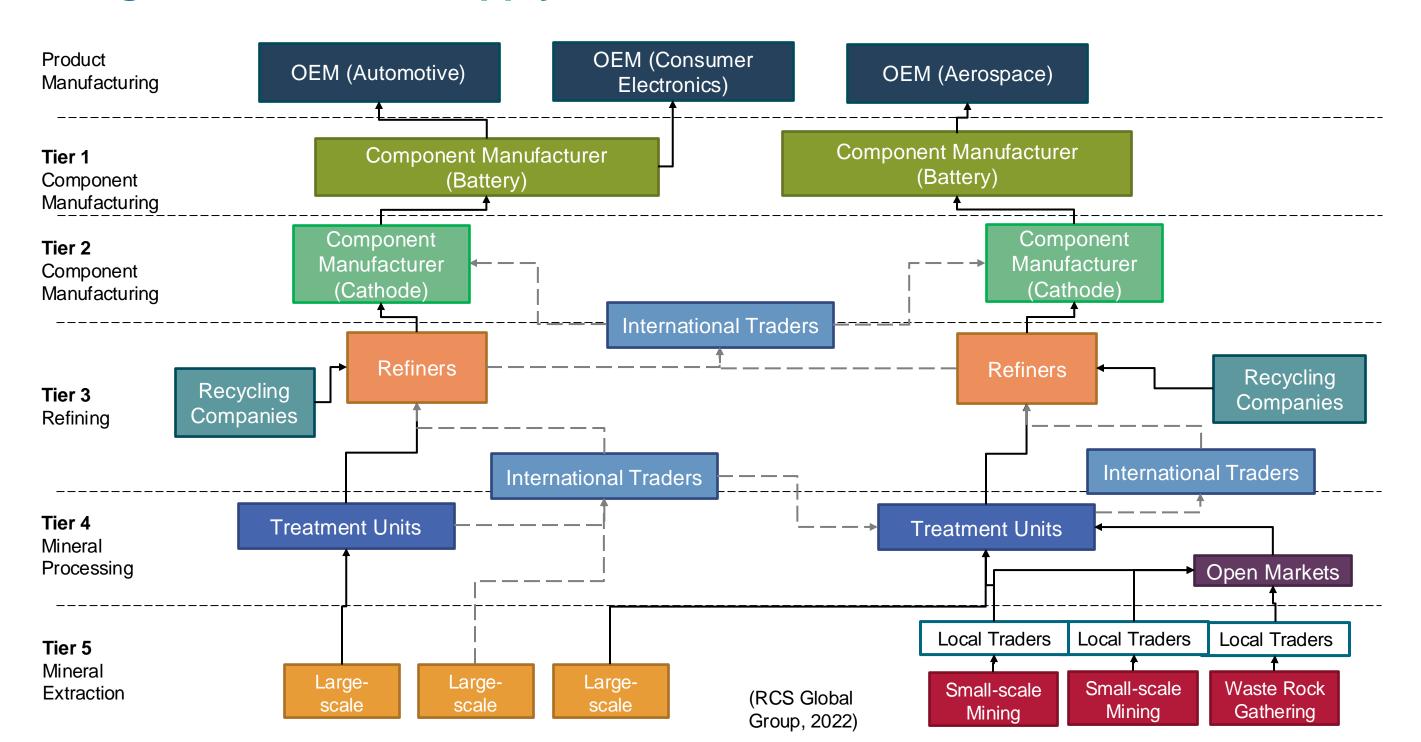
"A blind spot is an area in your range of vision that you cannot see properly but which you really should be able to see"

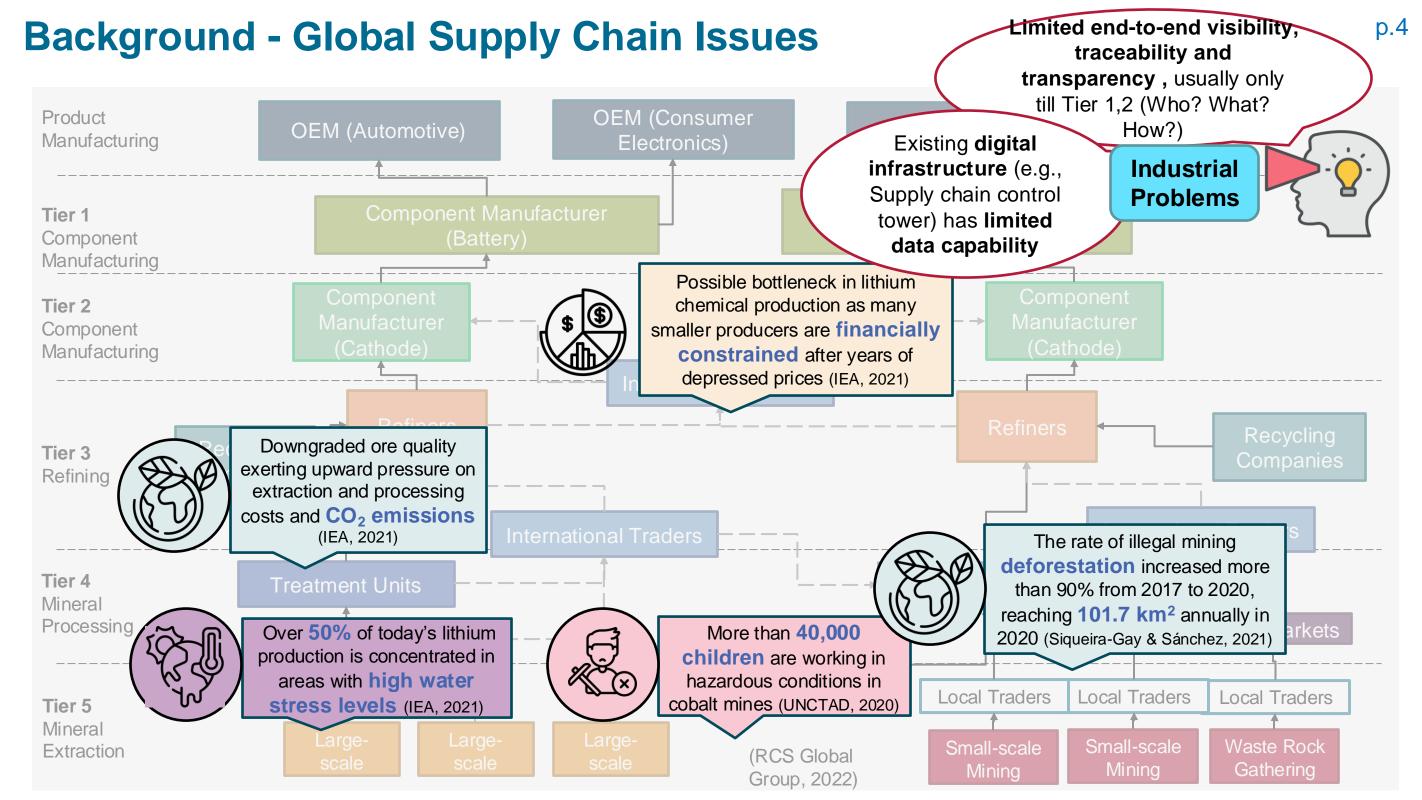
(Collins Dictionary, 2023)





Background - Global Supply Chain Issues





Background – Publicly Available Data & Big Data Analytics

Easily accessible data including:

- open data
- commercially available data
- data that can be obtained upon request

Big Data

Analytics

(de Assis Santos & Marques, 2022)

- > public-domain data
- data with limited accessibility

(Cooper & Coetzee, 2020)

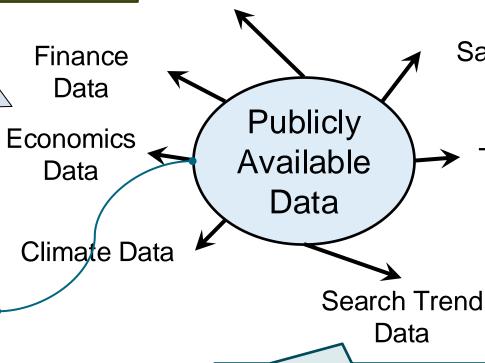
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Google Earth

Satellite Image

Trade Data

- Heterogeneous data types and formats
- High volume
- Generate supply chain knowledge



Spot illegal deforestation and forced labour in the Amazon rainforest (Andrews, 2022)

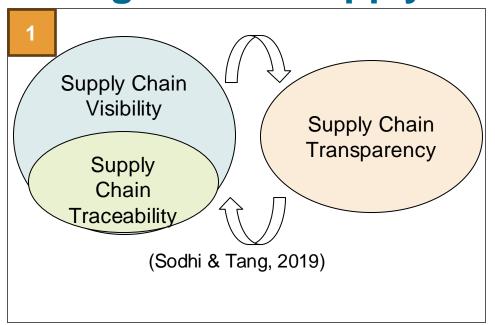
Use **import and export data** from UN Comtrade database for visualising the **geographical dispersion** of the commodity supply chain (Helbig et al., 2016)

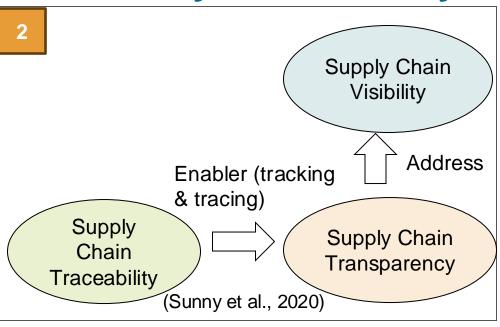
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Department of Engineering

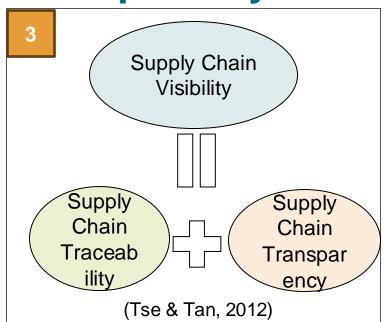
Use **Google Trends** data estimate product and service demand (Nikolopoulos et al., 2021)

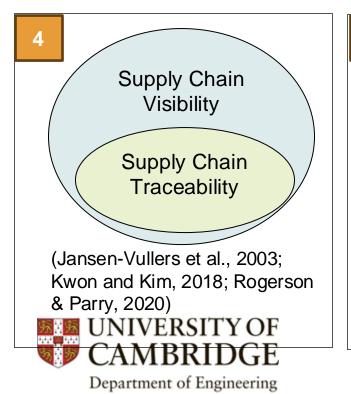


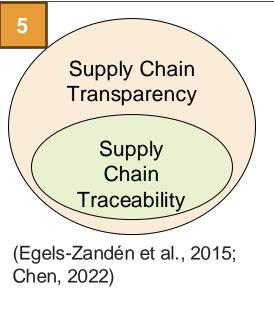
Background – Supply Chain Visibility, Traceability & Transparency

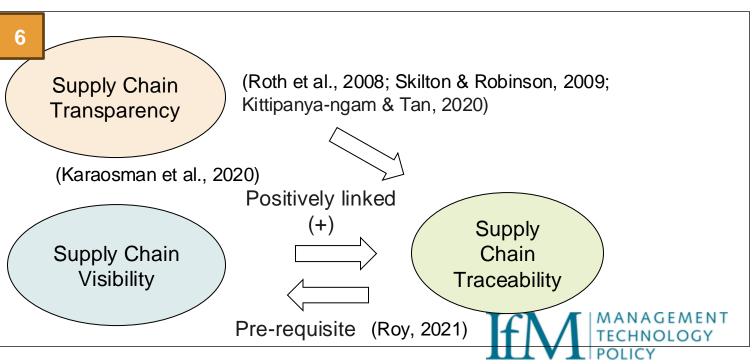




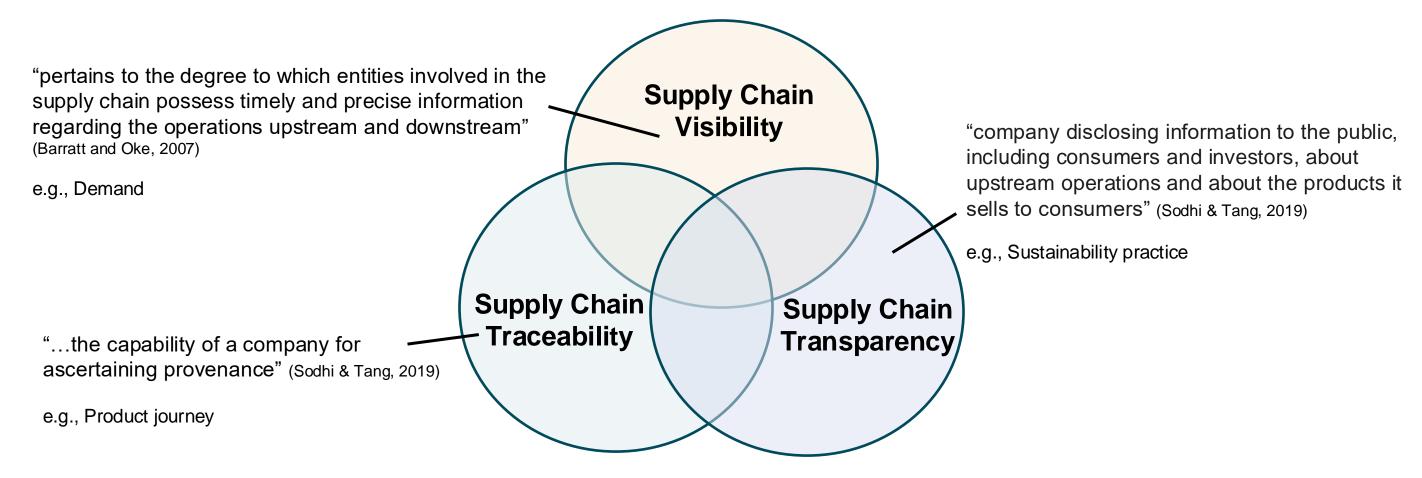








Background - Supply Chain Visibility, Traceability & Transparency



Considering data points related to supply chain nodes and the environments affecting them





Research Questions & Objectives

RQ: How can publicly available data be used with big data analytics (BDA) to develop the end-to-end supply chain visibility, traceability and transparency (SCVTT)?

Objectives

- Identify existing publicly available data that can be linked to SCVTT
- ☐ Understand the **exploitation** of publicly available data using BDA for the development of SCVTT, and their **challenges**





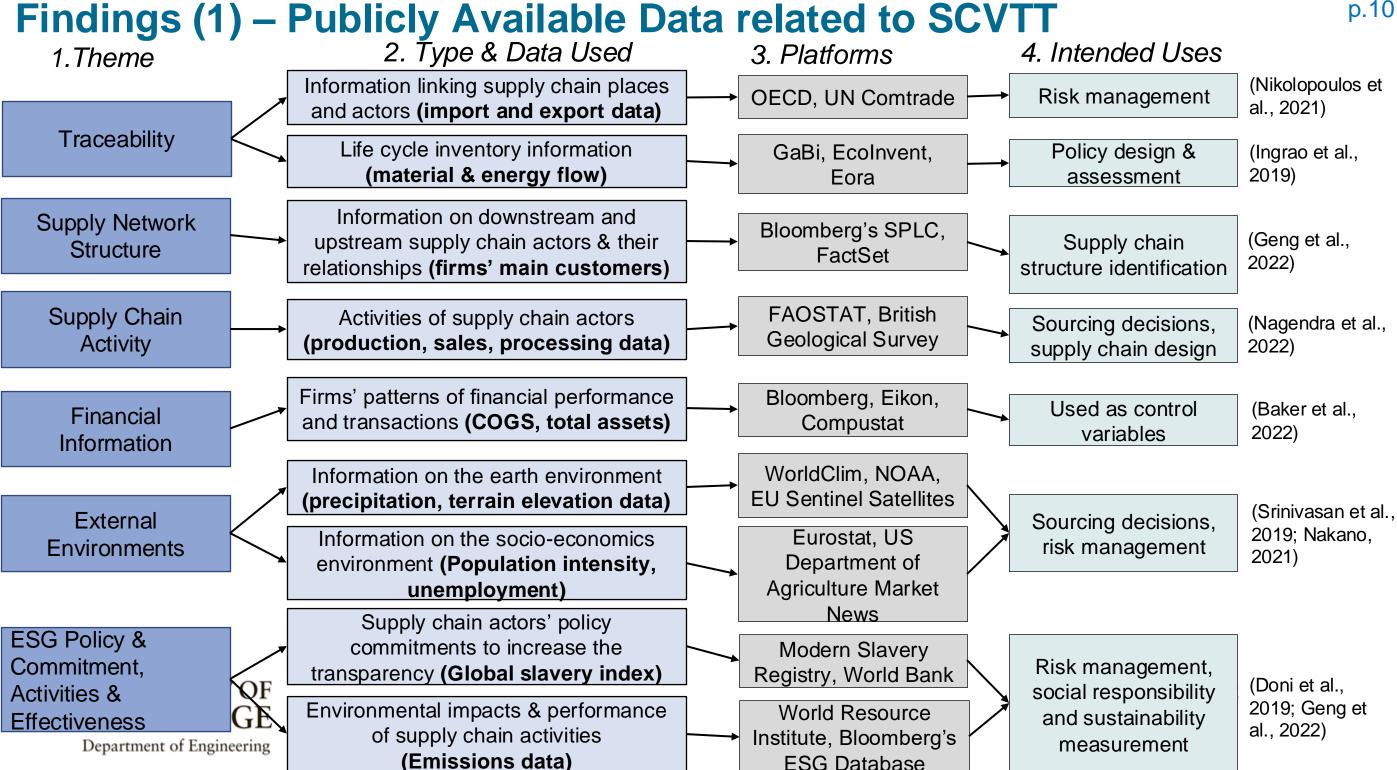
Research Method

A scoping review, 56 out of 2380 papers were selected for full review

Database	Web of Science				
Search field	Topic				
Search string	"supply chain*" AND ("simulation" OR "statistic*" OR "big data" OR "data-driven" OR "analytics" OR "machine learning" OR "data science" OR "data analy*" OR "operations research" OR "optimization" OR "optimisation" OR "open data" OR "public* data*" OR "database" OR "public* available data*" OR "data mining" OR "business intelligence")				
	Publication time: Last 5 years	Document type: Articles			
Filters	Language: English	Journal: Only ABS listed and with 3/4/4* scores			
	Inclusion criteria:				
Title, Abstract	•Application paper rather than review or conceptual notes				
and Full-text	•Applied in supply chain operation context				
screening	•Used publicly available data as the data source				
	•Applied any of the mentioned techniques/models				
	•Illustrated what and how publicly available data is used				
	Exclusion criteria:				
	•Studied the importance/capability/influence/culture of the mentioned techniques (especially using				
	some qualitative research methods like sur	rvey and interview)			







Findings (1) - Characteristics of Publicly Available Data related to SCVTT

p.11

Publicly available data used in existing research present great heterogeneity

Data Type

- Structured Data (Tabular data)
- Unstructured Data (Satellite image)

Data Provider

- Private providers (Bloomberg)
- Public institutions (OECD)
- NGOs (Transparency International)

Data Size

- Single data point (Port distance)
- Traditional data (Annual crop yield)
- Big data (Satellite data)

Data (File) Format

CSV, PDF, TIFF, JPEG, TXT, Webpage

Data Frequency

- Historical data (Crop production)
- Real-time data (Google Trends)
- Projected data (Weather forecasting)

Data Geographical Coverage

- Regional (State, Province)
- National
- Global

Data Unit Standard

- International System of Units
- British Imperial Units
- US Customary Units

Data Update Frequency

- (Near) Real-time
- Hourly, Daily, Monthly, Annual
- Static (Distance between ports)

Data Resolution

- Product level (Commodity price)
- Company level (Company's' financial performance)
- Sector level (Pollutant emission inventory)

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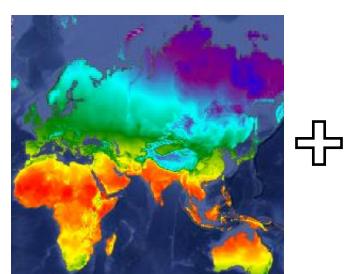
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(Ingrao et al., 2019; Srinivasan et al., 2019; van den Brink et al., 2020; Nikolopoulos et al., 2021; Taghizadeh et al., 2021; Baker et al., 2022; Geng et al., 2022; Nagendra et al., 2022; Ostroski et al., 2022)

Findings (1) – Publicly Available Data related to SCVTT

- o **Intentions** of using publicly available data as the data source are different:
 - (1) used as the **sole** data source (Baker et al., 2022)
 - (2) used to complement the primary data (Ingrao et al., 2019)
- o Integrating multiple publicly available data to generate meaningful information is a common practice (Srinivasan et al., 2019)
- o Data are from various disciplines and need to be **translated** into the supply chain contexts

Example: Using publicly available data for food sourcing decision making



	Optimal		Absolute		
	Min	Max	Min	Max	
Temperat. requir.	15	20	2	40	
Rainfall (annual)	500	1000	200	2000	
Latitude	30	-	55	70	
Altitude			-	4400	
Soil PH	6.5	7.5	6	8	

Candidate sourcing locations

Estimate the impact of climate change on food sourcing decisions

Historical & projected temperature and precipitation data from WorldClim database



Crop growth suitability data from FAO ECOCROP database

(Srinivasan et al., 2019)



Findings (2) – Exploitation of Publicly Available Data in Current Research

p.13

Data Source & Collection



Data Preprocessing



Data Storage & Management



Data Analysis & Use

Extracting data from

- Online database
- Reports
- Webpage
- Social media monitoring

Tools

- o API (Nagendra et al., 2020)
- Software, e.g., LCA Software
 GaBi (Suer et al., 2021)

Selection Reasons

- Desired data update frequency (Barker et al., 2022)
- Comprehensiveness (Dooley et al., 2019; Ingrao et al., 2019; Geng et al., 2022)
- Free (Costa et al., 2022)

Cleaning the Missing Data (data points or entire category)

- Remove the entry with missing data points (Geng et al., 2022)
- o Interpolation (Taghizadeh et al., 2021)
- Supplement with additional sources (Ostroski et al., 2022)

Big satellite data stored in **cloud platform** and fetched through API for analysis (Nagendra et al., 2020)

- Regression analysis (Barker et al., 2022)
- o Forecasting (Yu et al., 2019)
- Optimisation (Flores & Villalobos, 2020)
- O Simulation (Taghizadeh et al., 2021)
- Life cycle assessment (Ingrao et al., 2019)
- Multi-criteria decision making (Kougkoulos et al., 2021)

Summary

- Only 5 papers mentioned the data source selection reasons
- Data quality and reliability, data validation are rarely considered
- Mainly focus on the stages of data processing & analysis
- Data storage and management processes are missing from the reviewed research
- No standard set of criteria for analysis techniques/models selection with specific purpose





Findings (2) – Key Challenges of Exploiting Publicly Available Data

Data Source & Collection



Data Preprocessing



Data Storage & Management



Data Analysis & Use

Data definition

(precipitation vs rainfall)

Source selection

Data heterogeneity

(formats, resolutions, units, frequency, size)

Data

inconsistency

(e.g., measuring method)

Data accuracy, quality

&validity

Data source monitoring

(communication, different record/update frequency)

Missing data

Data conversion & fusion

Data triangulation

Data translation

Database model selection (SQL vs NoSQL)

> Storing highdimensional & heterogeneous data

Data analysis technique/ model selection

Results scalability

Computational time of large dataset





Discussions – Research gaps

Definitions & Requirements of SCVTT

- Differentiation between Visibility,Traceability & Transparency
- Agreed and standardised data requirements for SCVTT => What data is needed?

Use of publicly available data in SCVTT research

- Data definition and data source selection
- More attentions should be paid on data storage and management for sustained use
- A standardised framework to translate the data from various contexts to supply chain context

Supply Chain Visibility, Traceability & Transparency **Future** Research **Publicly Available Data Big Data** B Analytics

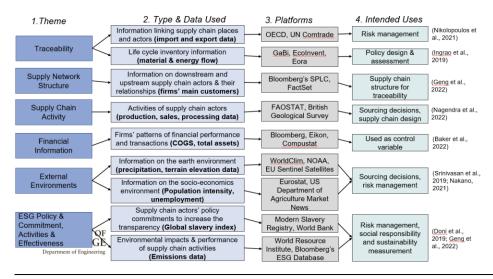
Implementation of big data analytics in research using publicly available data

Needs of empirical case which illustrates and guides the implementation of big data analytics (Constructs, Tools & Platforms)

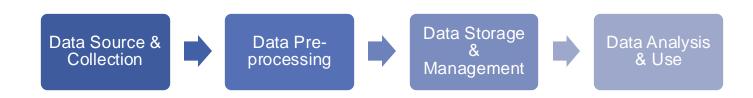


Conclusions

Objective 1: Identify existing publicly available data that can be linked to SCVTT



Objective 2: Understand the exploitation of publicly available data using BDA for the creation of SCVTT, and their challenges



Methods, tools, techniques, challenges

Contributions

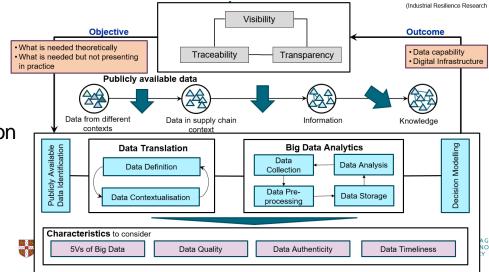
- Provide a holistic view of the use of publicly available data in current supply chain research
- Summarise themes related to SCVTT and corresponding data sources, increase the data capability of future research in the field of SCVTT & practical use
- Identify future research opportunities and propose a research framework



What's Next

What data is needed to develop SCVTT

- Data definition & translation
- Digital infrastructure development



References

Andrews, E. L. (2022, April 6). Detecting modern-day slavery from the Sky. Stanford Human-Centered Artificial Intelligence. https://hai.stanford.edu/news/detecting-modern-day-slavery-sky

Barker, J. M., Hofer, C., Hoberg, K., & Eroglu, C. (2022). Supplier inventory leanness and financial performance. Journal of Operations Management, 68(4), 385-407.

Barratt, M., & Barratt, R. (2011). Exploring internal and external supply chain linkages: Evidence from the field. Journal of Operations Management, 29(5), 514-528.

Barratt, M., & Oke, A. (2007). Antecedents of supply chain visibility in retail supply chains: a resource-based theory perspective. Journal of operations management, 25(6), 1217-1233.

Christopher, M., & Lee, H. (2004). Mitigating supply chain risk through improved confidence. International Journal of Physical Distribution & Logistics Management, 34(5), 388–396.

Cooper, A.K., Coetzee, S. (2020). On the Ethics of Using Publicly-Available Data. In: Hattingh, M., Matthee, M., Smuts, H., Pappas, I., Dwivedi, Y.K., Mäntymäki, M. (eds) Responsible Design, Implementation and Use of Information and Communication Technology. I3E 2020. Lecture Notes in Computer Science, vol 12067. Springer, Cham.

Costa, D., Quinteiro, P., Pereira, V., & Dias, A. C. (2022). Social life cycle assessment based on input-output analysis of the Portuguese pulp and paper sector. Journal of Cleaner Production, 330, 129851.

de Assis Santos, L. and Marques, L. (2022), "Big data analytics for supply chain risk management: research opportunities at process crossroads", Business Process Management Journal, Vol. 28 No. 4, pp. 1117-1145.

Dooley, K. J., Pathak, S. D., Kull, T. J., Wu, Z., Johnson, J., & Rabinovich, E. (2019). Process network modularity, commonality, and greenhouse gas emissions. Journal of Operations Management, 65(2), 93-113.

Flores, H., & Villalobos, J. R. (2020). A stochastic planning framework for the discovery of complementary, agricultural systems. European Journal of Operational Research, 280(2), 707-729.

Galbraith, J. (1973). Designing complex organizations. Reading, Mass.

Helbig, C., Gemechu, E. D., Pillain, B., Young, S. B., Thorenz, A., Tuma, A., & Sonnemann, G. (2016). Extending the geopolitical supply risk indicator: Application of life cycle sustainability assessment to the petrochemical supply chain of polyacrylonitrile-based carbon fibers. *Journal of Cleaner Production*, 137, 1170-1178.

IEA (2021), The Role of Critical Minerals in Clean Energy Transitions, IEA, Paris https://www.iea.org/reports/the-role-of-critical-minerals-in-clean-energy-transitions, License: CC BY 4.0

Kaipia, R., & Hartiala, H. (2006). Information-sharing in supply chains: five proposals on how to proceed. *The International Journal of Logistics Management*.

Kim, K. K., Ryoo, S. Y., & Jung, M. D. (2011). Inter-organizational information systems visibility in buyer-supplier relationships: The case of telecommunication equipment component manufacturing industry. Omega, 39(6), 667-676.

Kougkoulos, I., Cakir, M. S., Kunz, N., Boyd, D. S., Trautrims, A., Hatzinikolaou, K., & Gold, S. (2021). A multi-method approach to prioritize locations of labor exploitation for ground-based interventions. *Production and Operations Management*, 30(12), 4396-4411.

McCrea, B. (2005). EMS completes the visibility picture. Logistics Management, 44(6), 57-61.

Nagendra, N. P., Narayanamurthy, G., & Moser, R. (2020). Management of humanitarian relief operations using satellite big data analytics: The case of Kerala floods. Annals of operations research, 1-26.





References

Nikolopoulos, K., Punia, S., Schäfers, A., Tsinopoulos, C. and Vasilakis, C. (2021), "Forecasting and planning during a pandemic: COVID-19 growth rates, supply chain disruptions, and governmental decisions", *European Journal of Operational Research*, Vol. 290 No. 1, pp. 99–115.

Ostroski, A., Lagos, T., Prokopyev, O. A., & Khanna, V. (2022). Consumption-Based Accounting for Tracing Virtual Water Flows Associated with Beef Supply Chains in the United States. *Environmental Science & Technology*, 56(22), 16347-16356.

Pero, M., & Rossi, T. (2014). RFID technology for increasing visibility in ETO supply chains: a case study. *Production planning & control*, 25(11), 892-901.

RCS Global Group. (2022). COBALT SUPPLY CHAIN MAPPING REPORT.

Rogerson, M., & Parry, G. C. (2020). Blockchain: case studies in food supply chain visibility. Supply Chain Management: An International Journal, 25(5), 601-614.

Sodhi, M.S. and Tang, C.S. (2019), Research Opportunities in Supply Chain Transparency. Production and Operations Management, 28: 2946-2959. https://doi.org/10.1111/poms.13115

Srinivasan, R., Giannikas, V., Kumar, M., Guyot, R., & McFarlane, D. (2019). Modelling food sourcing decisions under climate change: A data-driven approach. Computers & Industrial Engineering, 128, 911-919.

Suer, J., Traverso, M., & Ahrenhold, F. (2021). Carbon footprint of scenarios towards climate-neutral steel according to ISO 14067. Journal of Cleaner Production, 318, 128588.

Sunny, J., Undralla, N., & Madhusudanan Pillai, V. (2020). Supply Chain Transparency through Blockchain-Based Traceability: An Overview with Demonstration. *Computers & Industrial Engineering*, 106895. doi:10.1016/j.cie.2020.106895

Taghizadeh, E., Venkatachalam, S., & Chinnam, R. B. (2021). Impact of deep-tier visibility on effective resilience assessment of supply networks. International Journal of Production Economics, 241, 108254.

UNCTAD. (2020). Commodities at a Glance: Special Issue on Strategic Battery Raw Materials. United Nations. https://www.un-ilibrary.org/content/books/9789210048293

Williams, B. D., Roh, J., Tokar, T., & Swink, M. (2013). Leveraging supply chain visibility for responsiveness: The moderating role of internal integration. Journal of operations management, 31(7-8), 543-554.

Yu, L., Zhao, Y., Tang, L., & Yang, Z. (2019). Online big data-driven oil consumption forecasting with Google trends. *International Journal of Forecasting*, 35(1), 213-223.





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Thank you!
Any questions or comments are welcome



