

How to translate economy into freight transportation?

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Knowledge for Tomorrow



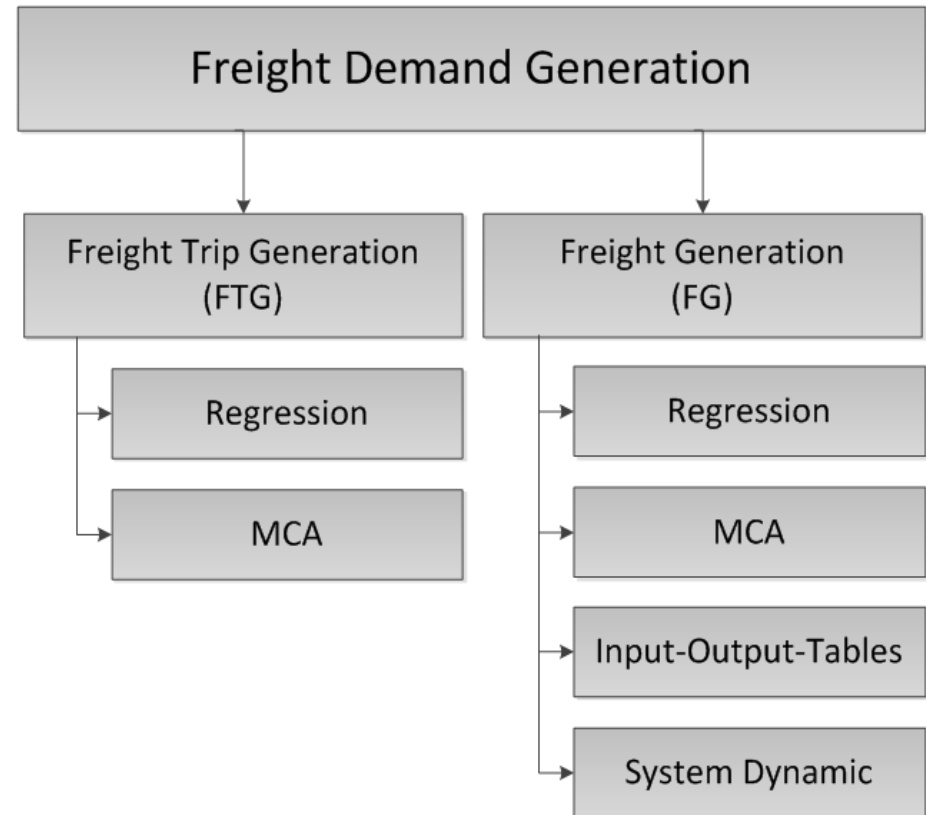
What is the challenge?

- Economy implies freight transport!
 - How much?
 - How much of which specific goods?
 - How much of which specific goods by which economic activity?



Who is mainly addressed by this challenge?

- Freight demand modelling
- Not freight trip generation



see Holguin-Veras et al. (2011): Freight generation, freight trip generation, and the perils of using constant trip rates



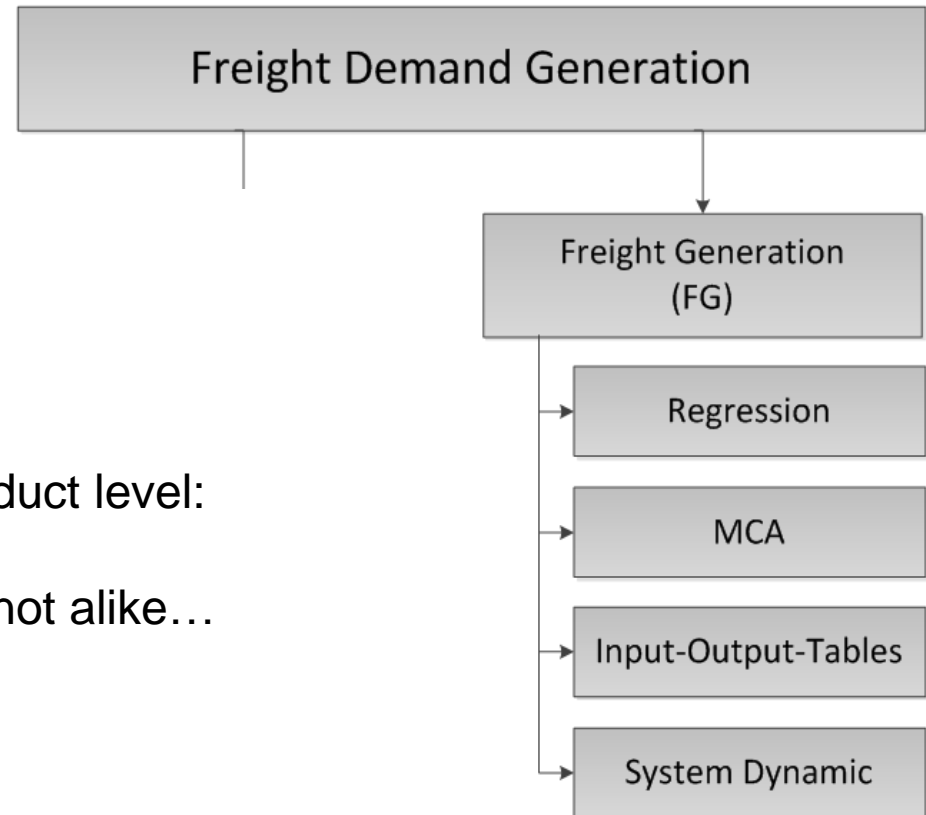
Who is mainly addressed by this challenge?

- Freight demand modelling
 - Not freight trip generation
 - But freight generation

- Ratio: value of good

- However, even on product level:

Two bottles of wine are not alike...



see Holguin-Veras et al. (2011): Freight generation, freight trip generation, and the perils of using constant trip rates



The idea is:

- Using disaggregated economic indicators to estimate freight generation
 1. Build a production and a consumption function (PF, CF) based on supply-use tables
 2. Derive weighting factors for PF and CF from supply-use tables
 3. Weight GVA and calculate the indicators for goods (CPA-classified)
 4. Transform CPA – goods into NSTR-24 classified goods

Perform a regression analysis



Step 1: production and consumption function

- Supply – Use – Table is the base
 - Supply tables containing producers prices
 - Use tables containing purchaser prices

		Industries (Nace)		
		1	...	59
Products (CPA)	1	€	€	€
	..	€	€	€
	59	€	€	€



Step 1: production and consumption function

- We utilize supply tables to extract weighted production function
 - Using the supply tables' information per row enables us to know which industries produce the same products.
- We utilize use tables to extract weighted consumption function
 - Using the use tables' information per row enables us to know which industries use the same products.

		Industries (Nace)		
		1	...	59
Products (CPA)	1	€	€	€
	..	€	€	€
	59	€	€	€



Step 1: production and consumption function

- We utilize supply tables to extract weighted production function
- We utilize use tables to extract weighted consumption function

$$\widehat{EI}_i = \sum_j (\alpha_{i,j} \cdot GVA_j)$$

\widehat{EI}_i : CPA classified economic indicator (€)
 i : index for products (CPA divisions)
 j : index for economic activities (NACE division)
 α : relevance of economic activity j for transportation of product i (for each option: use based, supply based, core industry based)
 $\sum_j \alpha_{i,j} = 1$ for each product i

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		1	...	59
Products (CPA)	1	€	€	€
	...	€	€	€
	59	€	€	€



Step 2: Derive weighting factors for PF and CF

- We utilize supply tables to extract weighted production function
- We utilize use tables to extract weighted consumption function

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Industries (Nace)

	1	...	59
1	€	€	€
..	€	€	€
59	€	€	€

$\alpha = \text{€ scaled to 1 per row.}$



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Industries (Nace)

		1	...	59
Products (CPA)	1	€	€	€
	...	€	€	€
	37	€	€	€

$\alpha = \text{€ scaled to 1 per row.}$



Step 3: Weight GVA and calculate the indicators

- We have the PF, CF and weighting factor by employing SUTs
- GVA from general economic statistics available
- Two economic indicators can be calculated now with PF and CF
 - 1 supply table based
 - 1 use table based



Step 3: Weight GVA and calculate the indicators

- We have the PF, CF and weighting factor by employing SUTs
- GVA from general economic statistics available
- Two economic indicators can be calculated now with PF and CF
 - 1 supply table based
 - 1 use table based
- However CPA classified → we intend a NSTR classified indicator
 - CPA are products in Euro
 - NSTR are transported commodities in tons
- We need a bridge matrix



Step 4: Transform CPA – into NSTR-24

- We need a bridge matrix



Statistisches Bundesamt (2008),
Amtsblatt der Europäischen Gemeinschaften (1998),
WIFO (2010),
TRAFICO et al. (2009),
Eurostat (2014) and
STATISTIK AUSTRIA (2014)]

Picture source: blog.archisnapper.com, adapted



Step 4: Transform CPA –into NSTR-24

- We need a bridge matrix to re-classify CPA into NSTR

NSTR24	CPA	β	NSTR24	CPA	β	NSTR24	CPA	β
01	01	0.33	13	27	0.51	24	01	0.1
02	01	0.36	14	26	0.88	24	05	0.2
03	01	0.12	15	14	1	24	12	1
03	05	0.34	16	24	0.09	24	15	0.1
04	02	1	16	25	0.06	24	16	0.8
04	20	1	17	24	0.01	24	17	0.3
05	17	0.07	17	25	0.01	24	18	0.3
05	18	0.07	18	24	0.85	24	19	0.3
05	19	0.07	18	25	0.59	24	21	0.2
05	36	0.06	19	21	0.8	24	22	1
05	37	0.07	20	29	0.8	24	24	0.05
06	15	0.9	20	30	0.33	24	25	0.34
06	16	0.2	20	31	0.7	24	26	0.05
07	01	0.09	20	32	0.33	24	27	0.05
07	05	0.46	20	33	0.33	24	28	0.1
08	10	1	20	34	0.9	24	29	0.2
09	11	0.01	20	35	0.9	24	30	0.67
09	23	0.01	21	28	0.22	24	31	0.3
10	11	0.99	21	27	0.16	24	32	0.67
10	23	0.99	22	26	0.07	24	33	0.67
11	13	0.92	23	17	0.63	24	34	0.1
11	27	0.25	23	18	0.63	24	35	0.1
12	13	0.08	23	19	0.63	24	36	0.34
12	27	0.03	23	36	0.6	24	37	0.25
13	28	0.68	23	37	0.68			



Step 4: Transform CPA –into NSTR-24

- We need a bridge matrix

NSTR24	CPA	β	NSTR24	CPA	β	NSTR24	CPA	β
01	01	0.33	13	27	0.51	24	01	0.1
02	01	0.36	14	26	0.88	24	05	0.2
03	01	0.12	15	14	1	24	12	1
03	05	0.34	16	24	0.09	24	15	0.1
04	02	1	16	25	0.06	24	16	0.8
04	20	1	17	24	0.01	24	17	0.3
05	17							0.3
05	18							0.3
05	19							0.2
05	36							1
05	37							0.05
06	15							0.34
06	16							0.05
07	01							0.05
07	05							0.1
08	10	1	20	34	0.9	24	29	0.2
09	11	0.01	20	35	0.9	24	30	0.67
09	23	0.01	21	28	0.22	24	31	0.3
10	11	0.99	21	27	0.16	24	32	0.67
10	23	0.99	22	26	0.07	24	33	0.67
11	13	0.92	23	17	0.63	24	34	0.1
11	27	0.25	23	18	0.63	24	35	0.1
12	13	0.08	23	19	0.63	24	36	0.34
12	27	0.03	23	36	0.6	24	37	0.25
13	28	0.68	23	37	0.68			

$$EI_{13,y} = \widehat{EI}_{27,y} \cdot 0.51 + \widehat{EI}_{28,y} \cdot 0.68$$

EI: economic indicator (€)

\widehat{EI} : CPA classified economic indicator (€)

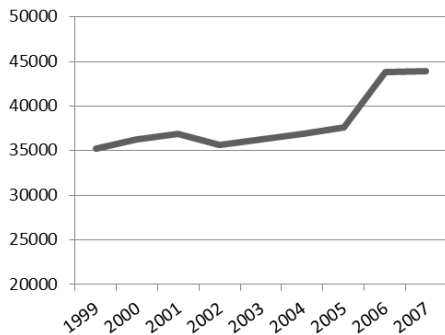
y: year



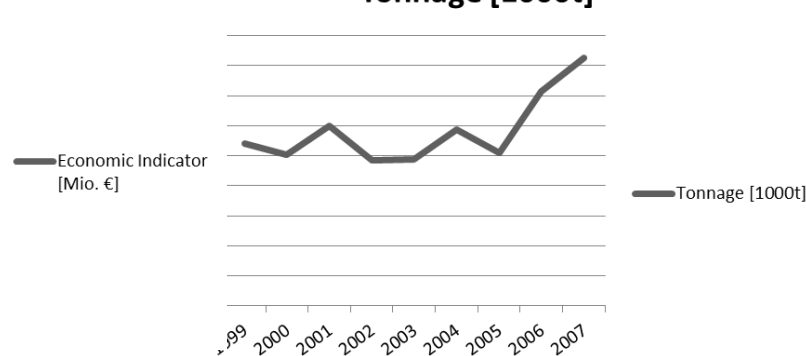
Finally: perform a regression analysis

- Now a regression analysis between the economic indicator and the transported tonnage is possible [data available from 1999-2007]
- Example NSTR-24 (13): Metal Products

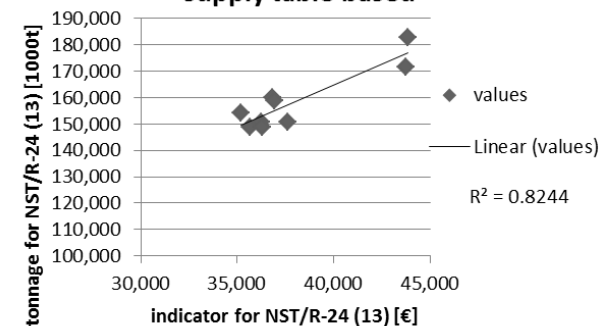
Economic Indicator [Mio. €]



Tonnage [1000t]

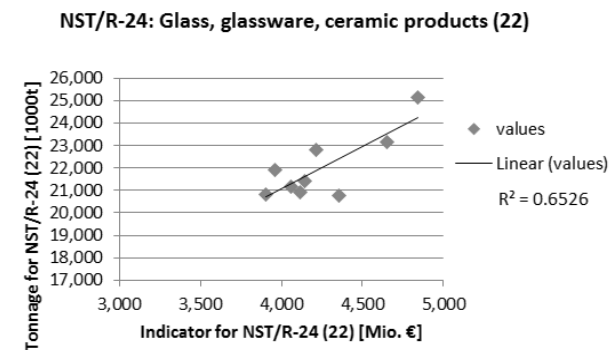
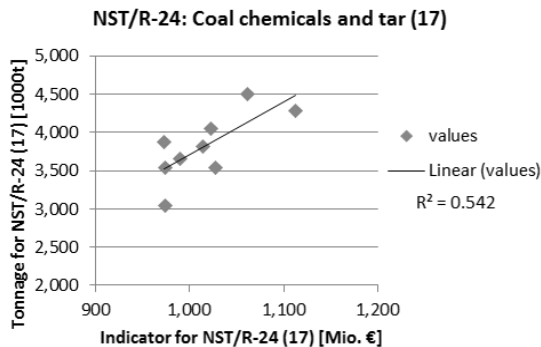
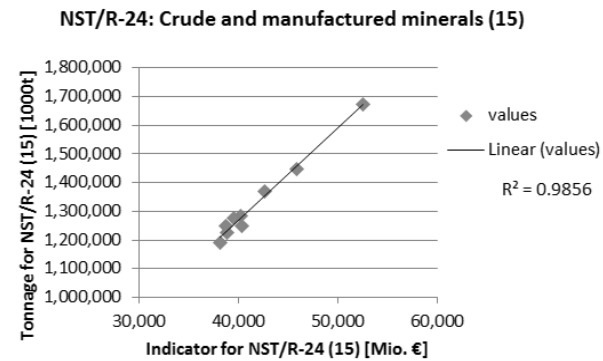
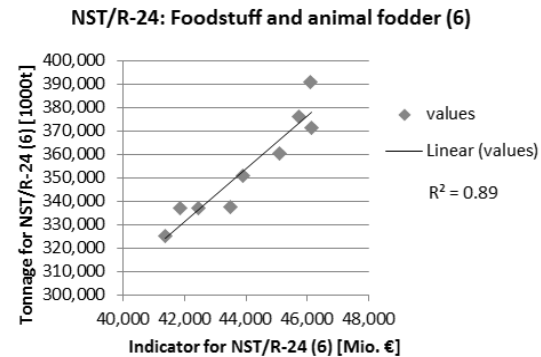


**Correlation indicator with tonnage for
NST/R-24 (13);
supply table based**



Finally: Result evaluation

- Now a regression analysis between the economic indicator and the transported tonnage is possible [data available from 1999-2007]



Result of the regression analysis

- Strong correlation is found for 11/24
- These 11 commodities represent 75% of the Tonnage in Germany (2007)
- Use-table based EI works slightly better 7/11

kind of good (NST/R-24)	R ² supply table based	R ² use table based	importance in Germany (share of tons in 2007)
1 Cereals	0.00	0.30	1.0%
2 Potatoes, other fresh or frozen fruits and vegetables	0.07	0.01	0.9%
3 Live animals, sugar beet	0.23	0.36	0.6%
4 Wood and cork	0.05	0.16	2.6%
5 Textiles, textile articles and man-made fibres, other raw animal and vegetable materials	0.07	0.09	0.5%
6 Foodstuff and animal fodder	0.15	0.89	10.2%
7 Oil seeds and oleaginous fruits and fats	0.70	0.67	0.7%
8 Solid minerals fuels	0.37	0.09	2.7%
9 Crude petroleum	0.33	0.03	0.0%
10 Petroleum products	0.10	0.48	5.0%
11 Iron ore, iron and steel waste and blast furnace dust	0.00	0.05	2.6%
12 Non-ferrous ores and waste	0.03	0.13	0.3%
13 Metal products	0.82	0.83	4.8%
14 Cement, lime, manufactured building materials	0.84	0.86	5.1%
15 Crude and manufactured minerals	0.48	0.99	33.4%
16 Natural and chemical fertilizers	0.27	0.43	1.0%
17 Coal chemicals, tar	0.46	0.54	0.1%
18 Chemicals other than coal chemicals and tar	0.18	0.34	6.7%
19 Paper pulp and waste paper	0.02	0.21	1.0%
20 Transport equipment, machinery, apparatus, engines, whether or not assembled, and parts thereof	0.97	0.87	4.0%
21 Manufactures of metal	0.79	0.83	1.5%
22 Glass, glassware, ceramic products	0.56	0.65	0.5%
23 Leather, textile, clothing, other manufactured articles	0.90	0.53	4.9%
24 Miscellaneous articles	0.92	0.83	9.8%
	explained tonnage		75.0%



Discussion of the method

- The key to translate economy into freight transportation is:
disaggregated economic indicator involving all industries
- Correlation is found
 - In case of Germany: 75% of the amount of goods is correlated
 - Useful in modeling issues: disaggregated goods
 - Goods forecasts possible with time dependent value densities
- Just public available data are used (EUROSTAT)
- Correlation is better than with the core industry



Discussion of the m

- The key to translate econor
disaggregated econ

- Correlation is found
- In case of Germany: 7!
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7 Oil seeds and oleaginous fruits and fats	0.70	0.67	0.70
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12 Non-ferrous ores and waste	0.03	0.13	0.05
13 Metal products	0.82	0.83	0.81
14 Cement, lime, manufactured building materials	0.84	0.86	0.78
15 Crude and manufactured minerals	0.48	0.99	no GVA data
16 Natural and chemical fertilizers	0.27	0.43	0.27
17 Coal chemicals, tar	0.46	0.54	0.46
18 Chemicals other than coal chemicals and tar	0.18	0.34	0.18
19 Paper pulp and waste paper	0.02	0.21	0.05
20 Transport equipment, machinery, apparatus, engines, whether or not assembled, and parts thereof	0.97	0.87	0.97
21 Manufactures of metal	0.79	0.83	0.81
22 Glass, glassware, ceramic products	0.56	0.65	0.49
23 Leather, textile, clothing, other manufactured articles	0.90	0.53	0.12
24 Miscellaneous articles	0.92	0.83	0.91



Discussion of the method

- We cannot derive a pattern why some goods are explained and others are not
 - Natural resources are more expensive when they become scarce
 - Price decrease/increase for commodities over years
 - The bridge matrix, which converts CPA-classified goods into NST/R-24-classified goods, is a great uncertainty
 - Taking into account the handling in the transport of goods
- Logistics modules needs new formulation
- Classification change (NST2007, NACE rev. 2)



Next steps

- We discussed the results but ultimately we see importance in giving further attention to the reasons why some product groups can still be explained while others cannot.
- Time-dependent value densities
- test the methodology for other European countries
- a regional investigation, on federal state or NUTS 3 level would also be desirable. Especially for SCGE or MRIO modelling needs and for European wide model needs is this investigation reasonable
- Test the methodology for the CPA classified EI



Vielen Dank für Ihre Aufmerksamkeit!
(Thank you for your attention)

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