

# Uncovering the Determinants of Shippers' Willingness to Shift from Road to Rail Freight Transport

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All pictures generated using Midjourney

# Outline

- Introduction
- Modal Shift
- Binary Choice Model
- Results & Discussion



# Introduction

Freight transport plays a significant role in Australia's economy.

Total Freight



Total Freight to grow 26%  
by 2050

Road Freight



Road Freight to increase 77% by  
2050

Rail Freight



Rail Freight projected to grow  
6% by 2050

# Modal Shift

Road is predominantly the mode of choice for urban, inter-urban and regional freight and most import supply chains.

- Heavy vehicle tax reforms
- Carbon emission tax




The need to encourage a Modal shift of freight from road to rail

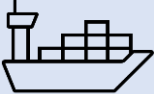



A change in the design of freight transport-related policies towards modal shift


# Data:


 Revealed Preference (RP) data collected in 2017


 Import and export container movements

 Covers 43,560 individual freight movements


 Data includes:

 Commodity type

 Commodity Weight

 Origin and Destination

 Transport mode choice

 Shipment Value (AUD)

 O-D Postcode & Timestamp



# Choice Modeling:

## When, Why and How Should Freight be Moved?

Optimization models  
such as linear  
programming



Choice models, such as  
binary choice models

# Binary Choice Model

$$P_n(i) = \frac{\exp(V_{in})}{\sum_{j=1}^k \exp(V_{jn})}$$

Where:

$P_n(i)$  = probability of shipper n choosing mode I,

$V_{jn}$  = utility derived by individual n from mode j,

K = number of available modes of transportation

Hence, the utility by a shipper n from mode j,  $V_{jn}$ , is derived as a linear function of the explanatory variables as follows:

$$V_{jn} = \beta_{0j} + \beta_{1j}X_{1n} + \beta_{2j}X_{2n} + \dots + \beta_{nj}X_{qn}$$

Where:

$\beta_{0j}$  = Alternative Specific constant for mode j,

$\beta_{1j}, \beta_{2j}, \dots, \beta_{nj}$  = Coefficients associated with explanatory variables

$X_{1j}, X_{2j}, \dots, X_{nj}$  = Explanatory variables for shipper n

Q = number of explanatory variables included in the model

# Choice Modeling:

For our model, the utility by a shipper  $n$  from mode  $j$ ,  $V_{jn}$ , is derived as a linear function of the explanatory variables as follows:

$$P_n(i) = \frac{\exp(V_{in})}{\sum_{j=1}^k \exp(V_{jn})}$$

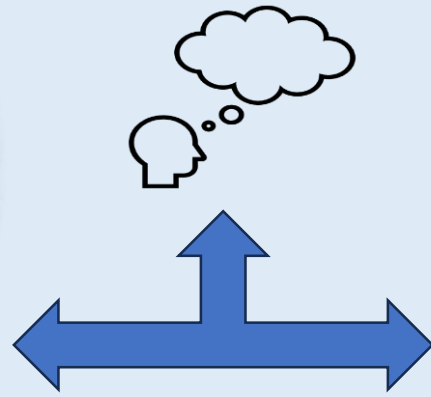
$$V_{rail} = \text{Constant}_{Rail} + \beta_0 + \text{Availability}_{Rail} + \beta_{Dist} * \text{Dist}_{sc} + \beta_1 * \text{Weight}_{agriculture} + \beta_2 * \text{Weight}_{mining} + \beta_3 * \text{Weight}_{agriculture} + \beta_4 * \text{Weight}_{chemical} + \beta_5 * \text{Weight}_{textile} + \beta_6 * \text{Weight}_{wood} + \beta_7 * \text{Weight}_{food} + \beta_8 * \text{Weight}_{household} + \beta_9 * \text{Weight}_{coal} + \beta_{10} * \text{Value}_{agriculture} + \beta_{11} * \text{Value}_{mining} + \beta_{12} * \text{Value}_{chemical} + \beta_{13} * \text{Value}_{textile} + \beta_{14} * \text{Value}_{wood} + \beta_{15} * \text{Value}_{household} + \beta_{16} * \text{Value}_{coal}$$



# Results

Variable	Value	Rob. Std. err	Rob p-value	Value	Rob. Std. err	Rob p-value
<b>Commodity type</b>	<b>Shipment Value</b>			<b>Shipment Weight</b>		
Agricultural products	10.30	1.05	0.00	0.02	0.01	0.00
chemical products	-2.89	0.71	0.00	0.07	0.07	0.01
textiles	0.00	0.00	0.00	0.04	0.01	0.00
craft products	-15.6	4.40	0.00	0.05	0.01	0.00
Household materials	0.95	0.18	0.00	0.00	0.00	0.00
construction materials	-3.94	0.91	0.00	0.07	0.01	0.00
Variable	Value	Rob. Std. err	Rob p-value			
Alternative Specific Constant of Rail	-13.70	1.57	0.00			
Distance	0.36	0.04	0.00			
Rail mode accessibility	9.45	1.56	0.00			
Statistics	Value					
No of parameters:	14					
Final log likelihood	-2680.88					
Akaike Information Criterion	5389.77					
Bayesian Information Criterion	5508.19					

# Discussion



Rail is the least preferred mode (ASC = -13.7)

# Discussion

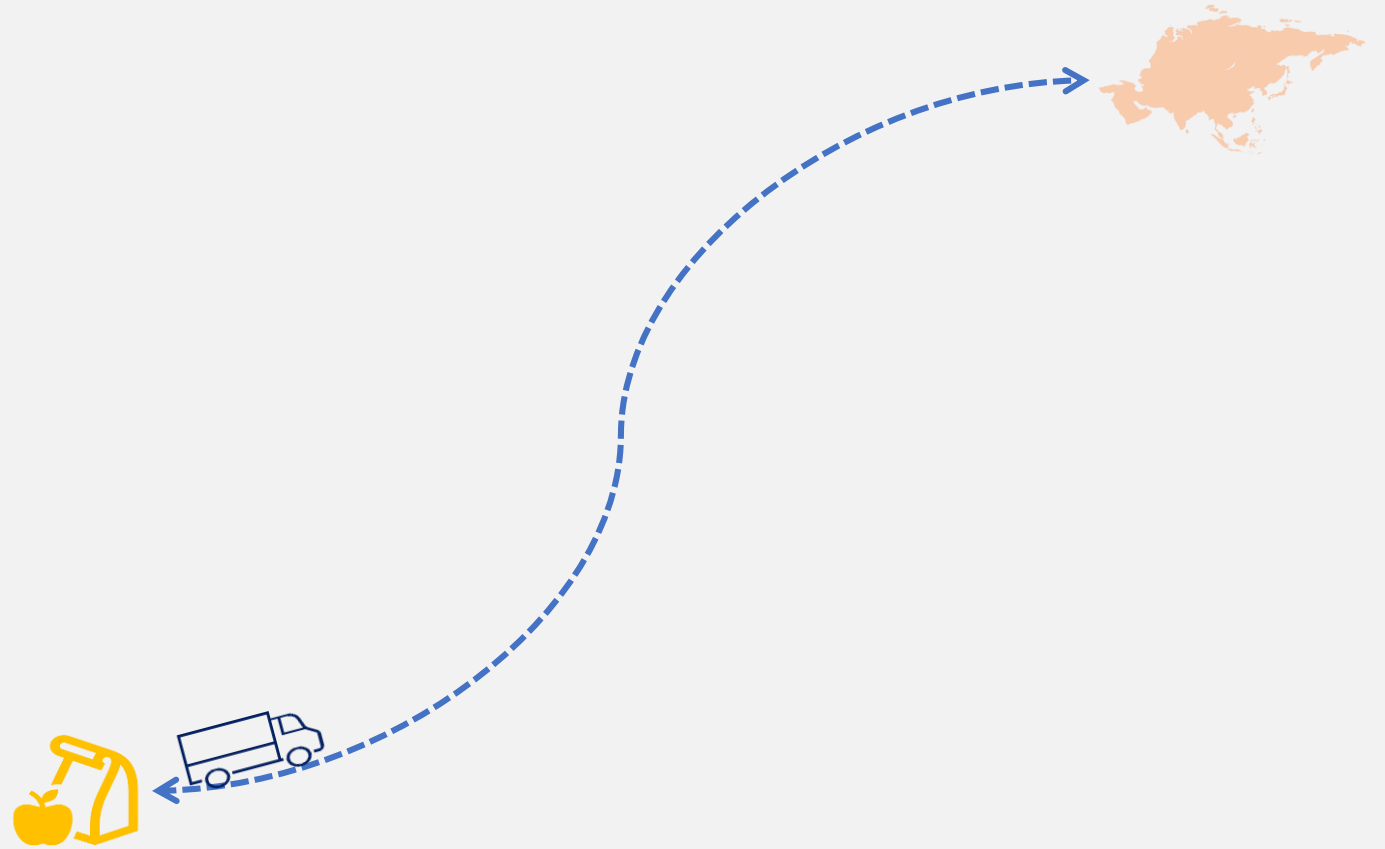
- ❑ Weight and shipment size influences mode choice significantly



For commodities such as food, dairy, fruit, and beverages, increasing the shipment size increases the likelihood of using rail  
(co-efficient = 0.0538)



# Discussion



Longer distance increases the probability of choosing rail

# Discussion

- ❑ AUD Value of the commodity plays a significant role (high-value commodities shippers choose road)



For agricultural, forestry and livestock products with a low shipment value (ship = rail)

For commodities with high shipment value, such as household items(ship = road)

# Discussion



Rail Availability

Mode accessibility & availability significantly influences choice (co-efficient = 9.45)

# Recommendations for Policymakers

- ❑ Encouraging modal shift isn't just about building infrastructure but creating connected modal networks
- ❑ Emphasis on improving the speed and reliability of rail transport e.g., speed freight trains
- ❑ Enabling freight bundling options by encouraging partnerships between shippers



THANK YOU





# Excluded Variables

The model some variables and estimates due to insignificant results:

- Empty container (Com 10)
  
- Freight values for:
  - Mining, coal, limestone, metallic ores, nonmetallic minerals (Com2)
  
  - Metallic and machinery products, primary and fabricated metal products, electronics, electrical machinery, transport equipment (Com3)
  
  - Light industrial products, textiles, leather (Com5)
  
  - Food, dairy, fruit, beverages, tobacco, seafood (Com7)

# Excluded Variables

The model some variables and estimates due to insignificant results:

- Freight weight for:
  - Agricultural, forestry, fishery and livestock products (Com1)
  - Metallic and machinery products, primary and fabricated metal products, electronics, electrical machinery, transport equipment (Com3)
  - Wood and paper products, lumber or wood products, pulp, paper or allied products, printed matter (Com6)
  - Coal products, rubber or plastic products, clay, concrete, glass, and stone products