

Modeling Consumer Preferences for Low-CO₂ Emissions Aircraft in China

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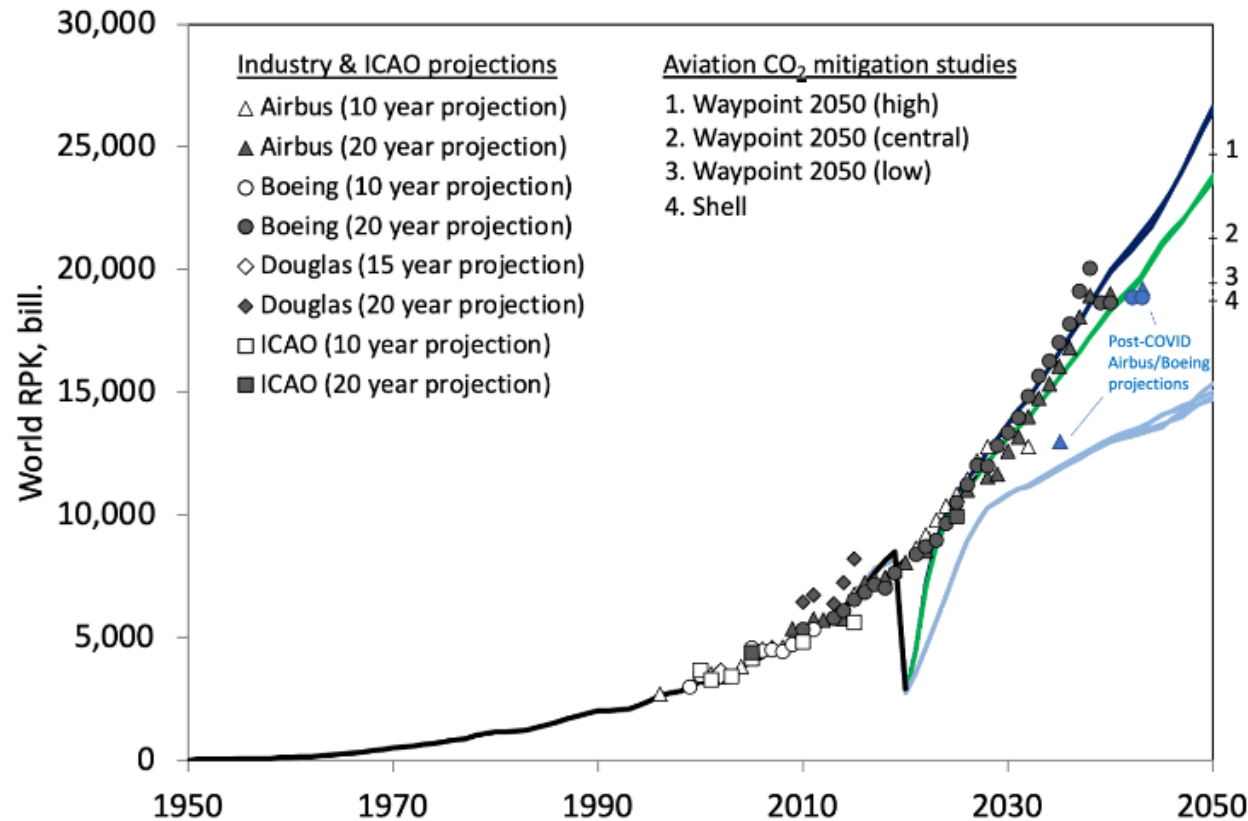
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Introduction



World revenue passenger-km travelled (RPK)

Aviation: Technologies and fuels to support climate ambitions towards 2050. *Concawe*. Report No. 5/23, 2023, p. 4.

- Global passenger aviation demand forecast to grow 4%/year despite COVID-19 impact
- In 2019, passenger flights contributed to ~ 85% of CO₂ emissions in commercial aviation (ICCT,2020)
- Aviation sector facing increasing pressure to mitigate CO₂ and non-CO₂ impact
- Achieving industry net-zero targets requires disruptive technological change

Introduction

Challenges

Achieving aviation net-zero goals will likely require an increase in airfares

Opportunities

Consumer's willingness to pay for green aviation

Social acceptance

- *Consumer attitudes*
- *Consumer choices*
- *Willingness-to-pay*



Airline behaviour

- *Airline choices*
- *Responses to new technologies*
- *Airline profitability*

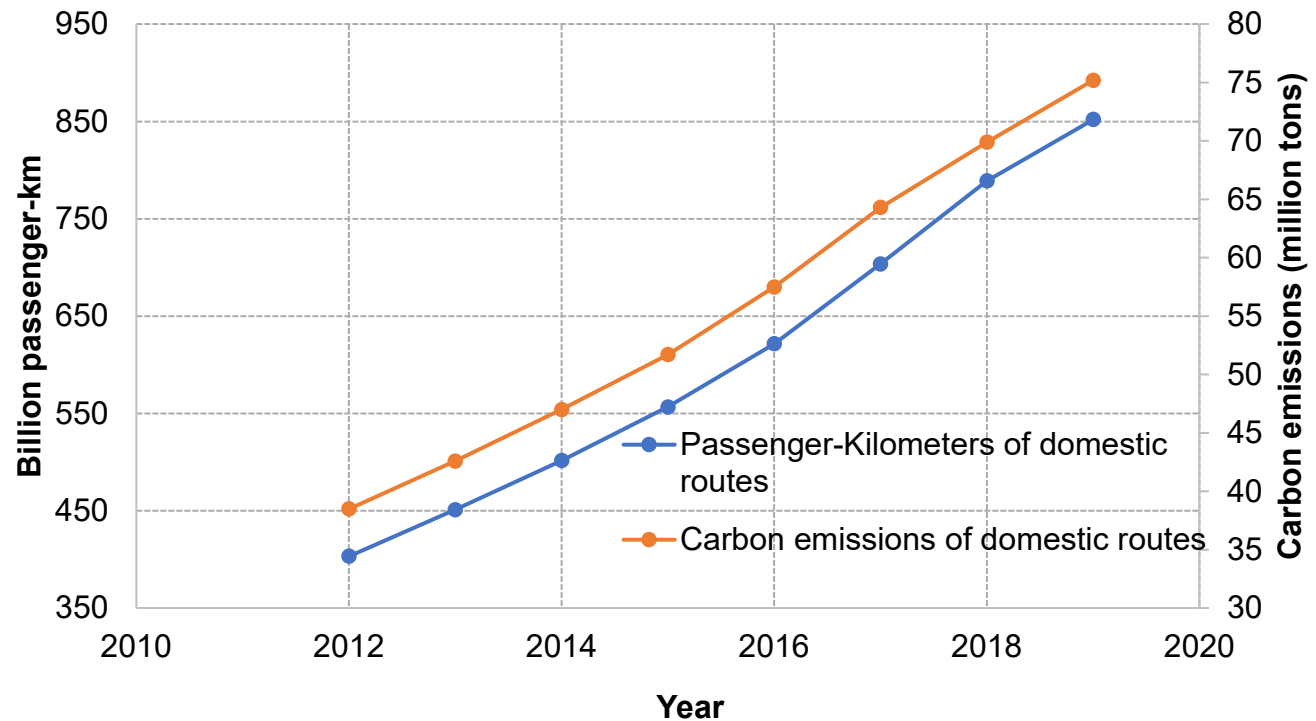
- Public opinion can either facilitate or impede the deployment of these disruptive technologies and fuels

Introduction

- Study focuses on Chinese domestic market to investigate consumer preferences for low-CO₂ emissions aircraft
- 2012 - 2019, domestic passenger air travel grew by 111%, related CO₂ emissions have increased by 95%

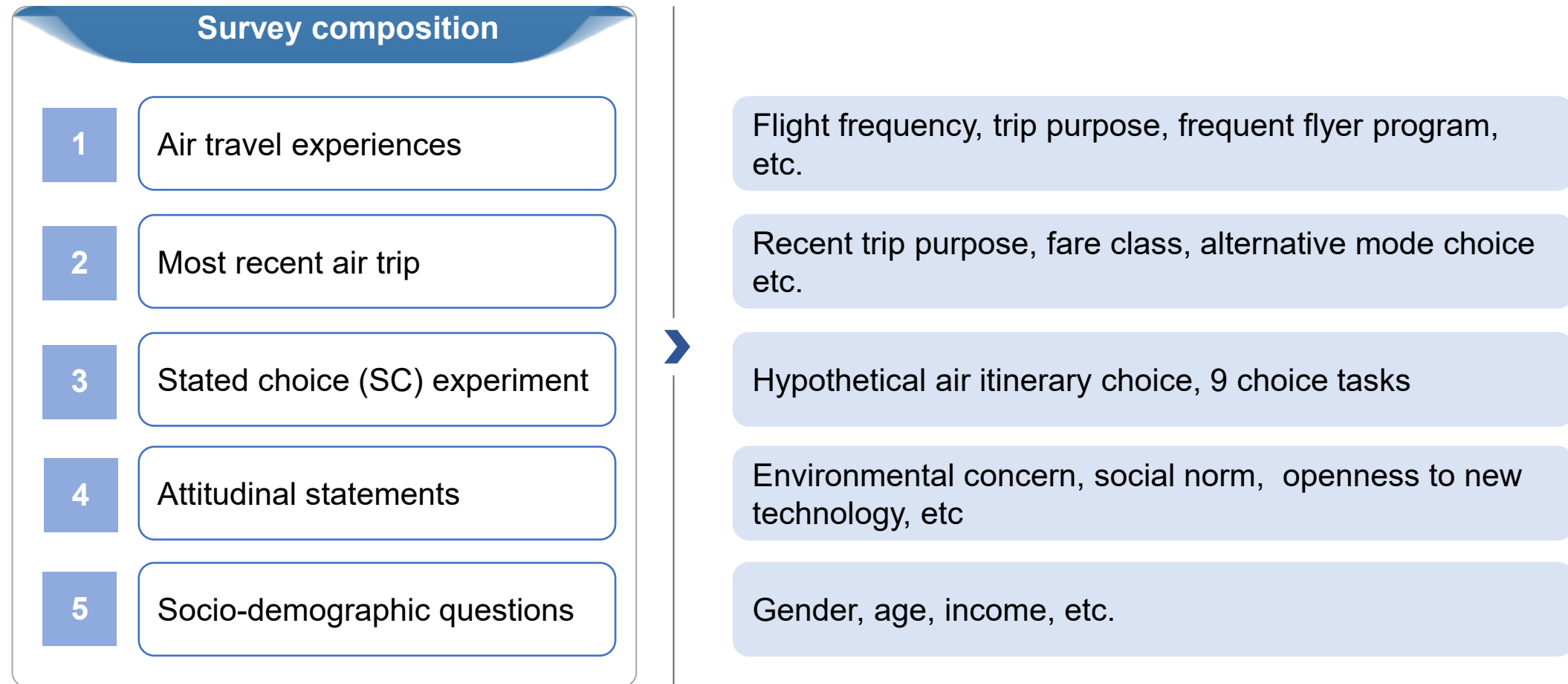
Research Objectives:

- Identify preference heterogeneity across individuals and market segments
- Analyse characteristics of potential early adopters
- Quantify consumer willingness to pay



Passenger-km Data: Civil Aviation Administration of China
CO₂ emissions of domestic routes adapted from (Cui et al., 2022)

Methodology - Stated Preference (SP) Survey



Stated Choice (SC) Experiment

Which flight ticket would you choose for a short-haul trip in China?

Note: Short-haul flights are defined as those less than 700 km (e.g., Beijing to Shenyang)

Sustainable Aviation Fuels (SAF) powered conventional aircraft

Zero-emission aircraft: all-electric aircraft and hydrogen powered aircraft

	Option 1	Option 2	Option 3
Aircraft Type	Hydrogen aircraft	SAF powered conventional aircraft	Conventional aircraft
Life Cycle CO2 Emissions Compared to The Conventional Aircraft	100 % reduction	60% reduction	0% reduction
Ticket Price (¥)	700	840	700
Flight Time (h)	2h15	1h30	1h30

Your choice:

Option 1

Option 2

Option 3

Methodology - Modelling Techniques

- Apply discrete choice modelling to estimate data from the SC experiment
- Assume utility maximisation behaviour
- The observed discrete choices are dependent variable
- Identify factors influencing choices:
 - Characteristics of choice options
 - Characteristics of decision-maker (e.g., age, income, etc.)
- Multinomial logit model (MNL): the probability of individual n choosing air itinerary option

$$P_{ni} = \frac{e^{V_{ni}}}{\sum_{j=1}^J e^{V_{nj}}} \text{ where } J \text{ is the total number of air itinerary options}$$

Methodology - Modelling Techniques

- Base model

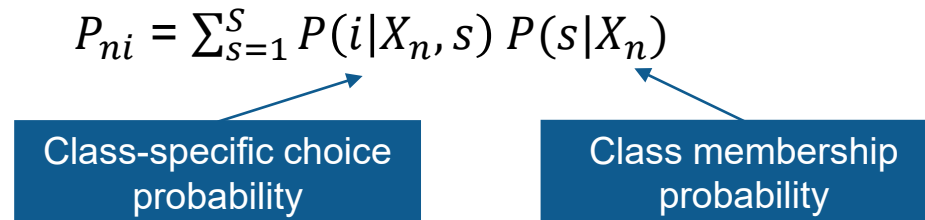
$$V_i = ASC_i + \beta_{fare} * fare + \beta_{flight\ time} * flight\ time + \beta_{aircraft\ type} * aircraft\ type + \beta_{emission\ reduction} * emission_reduction$$

- Incorporating systematic heterogeneity

$$V_i = ASC_i + (\beta_{fare} + \Delta_{\beta_{fare}} * business) * fare + (\beta_{flight\ time} + \Delta_{\beta_{flight\ time}} * business) * flight\ time \dots$$

- Latent class model

the probability of individual n choosing air itinerary option i :

$$P_{ni} = \sum_{s=1}^S P(i|X_n, s) P(s|X_n)$$


Class-specific choice probability

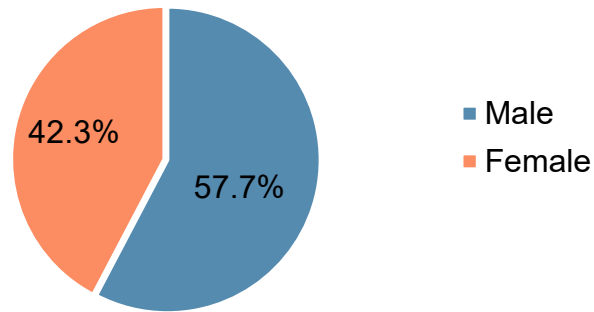
Class membership probability

where X_n is characteristics of itinerary option i , s is latent class

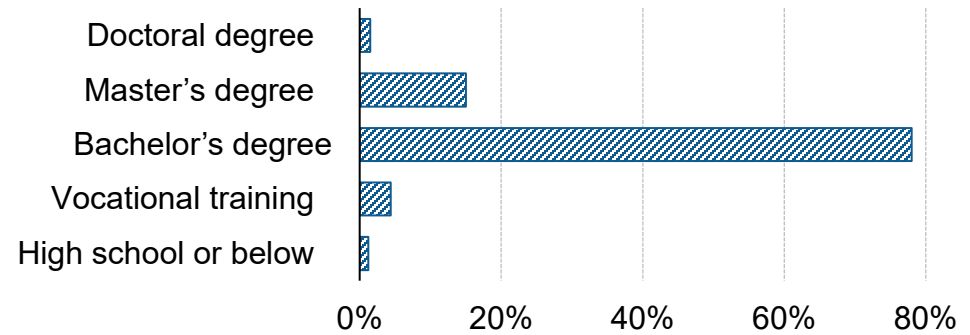
Data

Web-based survey: 3,187 respondents, who have flown before (28,683 SC observations)

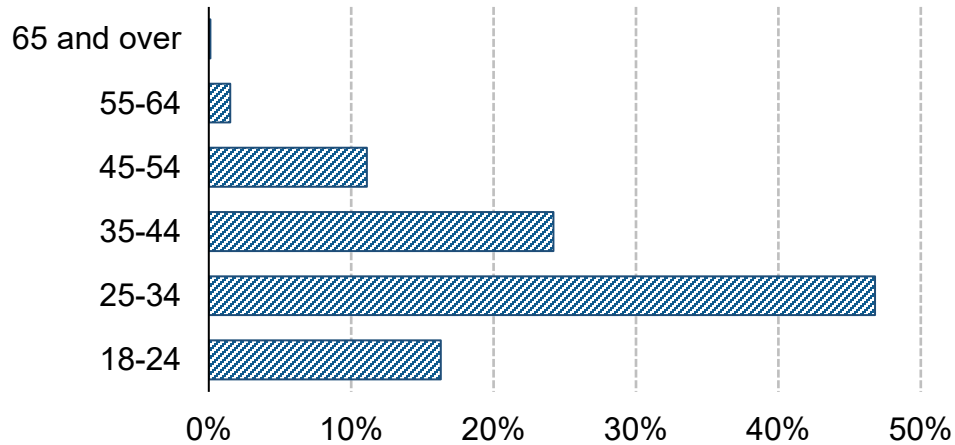
Gender



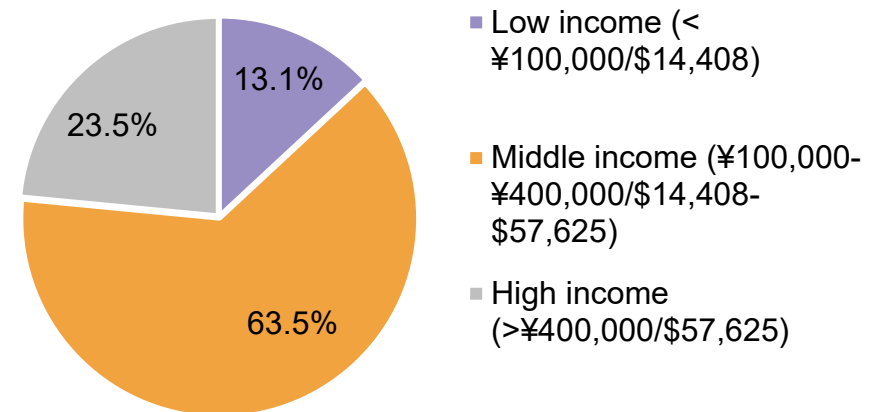
Education



Age

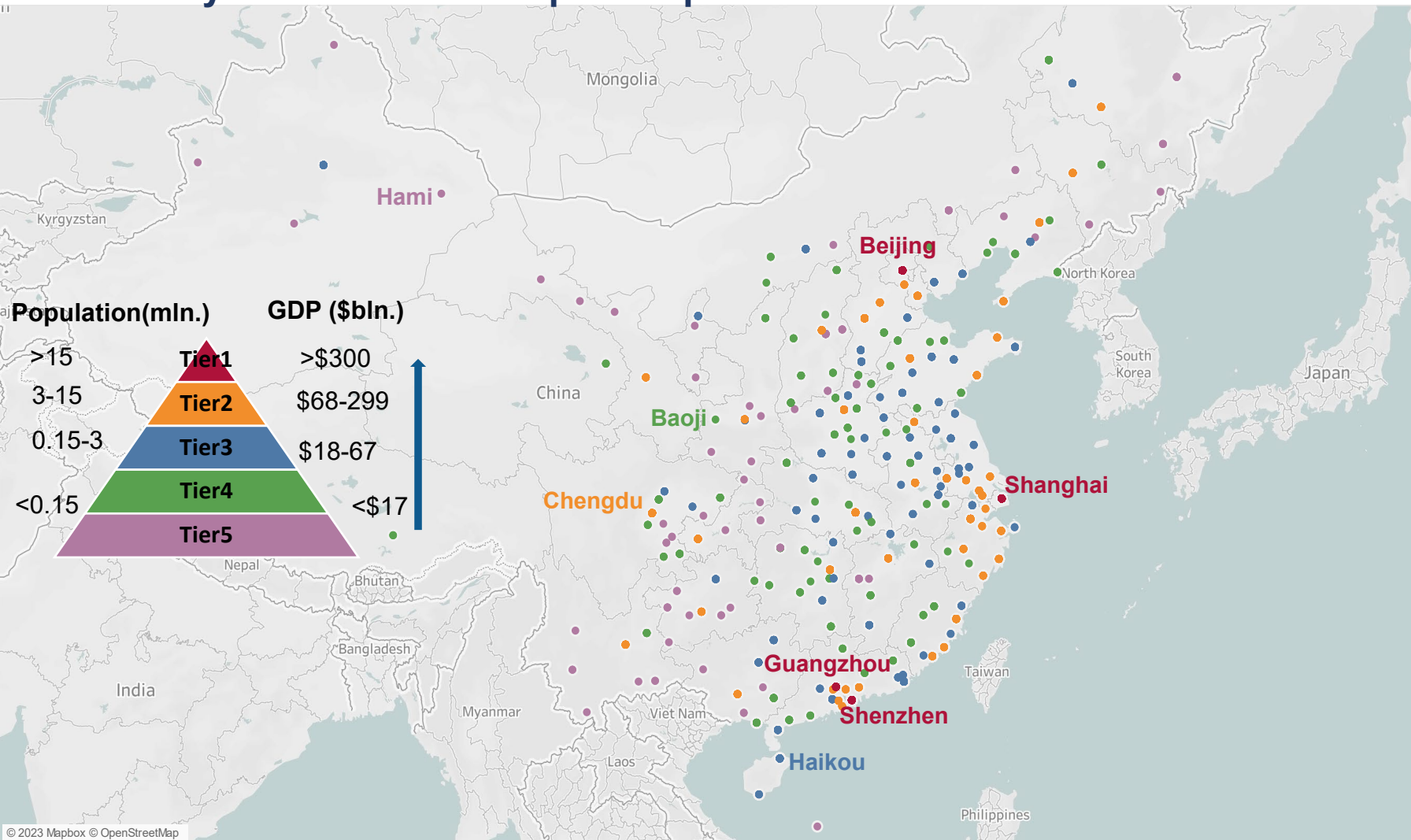


Household income



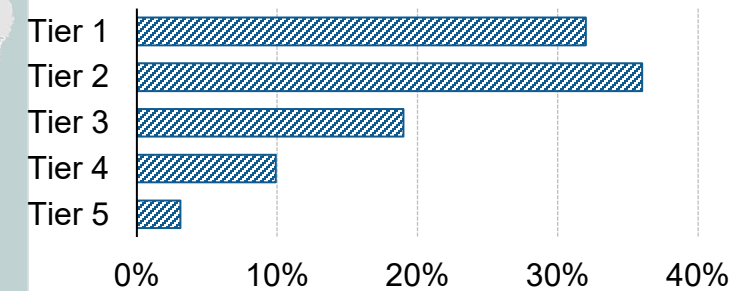
Data

City Locations of Sample Respondents



- Tier1: Beijing, Shanghai, Shenzhen, Guangzhou
- Tier2: e.g., Chengdu
- Tier3: e.g., Haikou
- Tier4: e.g., Baoji
- Tier5: e.g., Hami

City-tier



- Sample Coverage: All 31 Chinese administrative regions, with 63% residing in Tier 1 and Tier 2 cities

Estimation results

- Preference order (all respondents):
 Electric aircraft > Hydrogen aircraft > SAF-powered conventional aircraft > Conventional aircraft

 Corresponding WTP for alternative aircraft: \$89, \$63, \$44
- Preference order (business travelers):
 Hydrogen aircraft > SAF-powered conventional aircraft > Conventional aircraft

 Corresponding WTP for alternative aircraft: \$98, \$69
- Business travellers: Higher flight time sensitivity, lower fare sensitivity

	Panel effect model		Panel effect model with interaction	
Number of Parameters	9		15	
Number of Respondents	3187		3187	
Number of Observations	28683		28683	
LL (Log likelihood)	-27668.45		-26980.11	
Adj. Rho-squared	0.122		0.143	
AIC	55354.91		53990.22	
BIC	55429.28		54114.18	
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ASC1	0.1743	10.74	0.1587	9.61
ASC2	0.1412	8.49	0.1373	7.94
ASC3(base)				
Fare (¥)	-0.0014	-32.63	-0.0023	-21.39
Business travelers			0.0009	7.86
Flight time (min)	-0.0034	-4.41	-0.0025	-2.41
Business travelers			-0.0110	-6.91
SAF powered conventional aircraft	0.4119	11.22	0.2271	3.95
Business travelers			0.4345	5.79
Electric aircraft	0.8353	13.44	0.8455	9.06
Business travelers			0.1225	0.97
Hydrogen aircraft	0.5962	10.61	0.3800	4.46
Business travelers			0.5504	4.82
Conventional aircraft(base)				
Life-cycle CO2 emissions compared to the conventional aircraft (%)	0.0053	11.18	0.0062	8.36
Business travelers			0.0001	0.08
Sigma panel	0.1938	9.02	0.2121	10.11

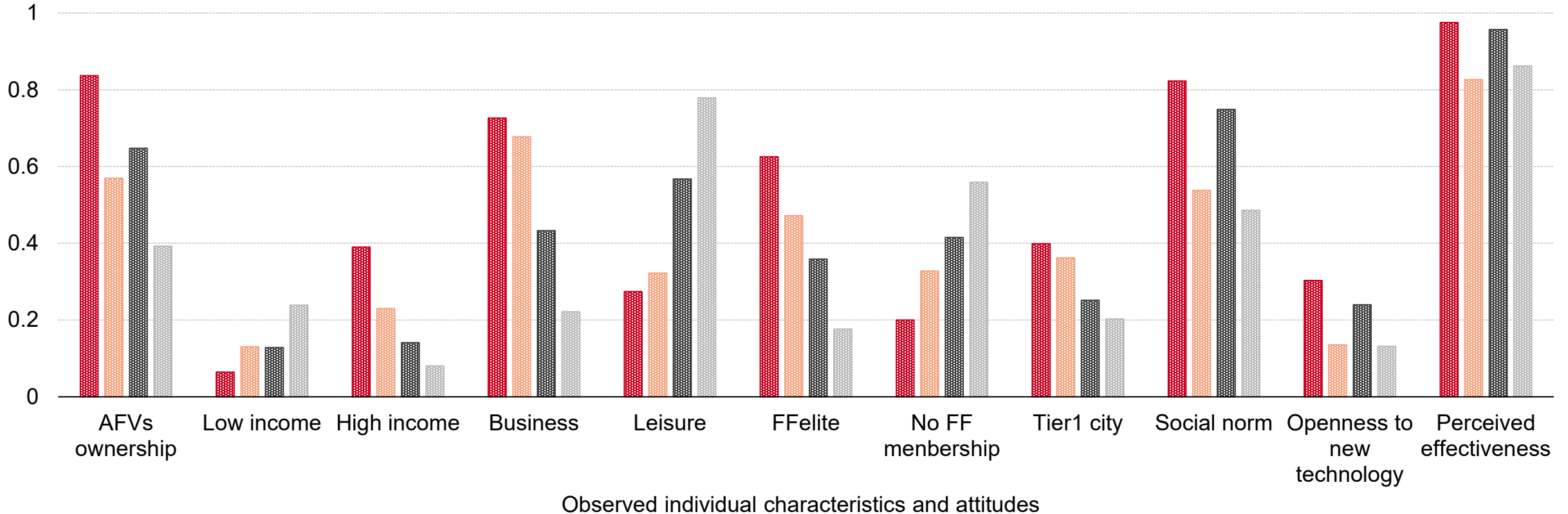
Estimation results

- Business travelers' interaction with employer-paying option
 - Employer-paying: Reduced fare sensitivity, higher value of time (VOT) (\$152 vs. \$27 self-paying)
 - Employer-covered business travelers have a greater WTP for flying on an alternative aircraft
- Four distinct groups were identified from the latent class model:
 - Class 1 (32%), Class 2 (29%), Class 3(18%), Class 4(21%)
 - Class 1 and Class 2: business travelers, similar VOT values (141\$/h vs. 140\$/h), Class 1 shows higher preference for alternative aircraft
 - Class 3 and Class 4: leisure travelers, similar VOT values (22\$/h vs. 23\$/h), Class 3 shows higher preference for alternative aircraft

Estimation results

Class membership probabilities

- Eco-Elite Business Travelers [Class 1] (32%)
- Traditional Business Travelers [Class 2] (29%)
- Green-Oriented Leisure Travelers [Class 3] (18%)
- Budget-Conscious Leisure Travelers [Class 4] (21%)



Conclusions

- Chinese respondents exhibit a preference for low-CO₂ emissions aircraft over conventional aircraft
- Identified four potential segments
 - Significant variations in WTP for flying on alternative aircraft across four segments:
\$36 ~ \$177 for SAF powered conventional aircraft, \$13 ~ \$208 for electric aircraft, \$9 ~ \$219 for hydrogen aircraft
- Early adopters are likely to be affluent individuals who have prior experience with AFVs. They tend to embrace new technologies, believe in the positive impact of adopting low-CO₂ emissions aircraft, and are susceptible to social influence
- Limitations include selection bias, hypothetical bias due to SP data
- Establishes groundwork for future research into interactions between airline decision making and consumer preferences in emerging markets

More information:

<http://www.atslab.org/>