# Impact of Regional Coupons on Tourist Consumption Behavior

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• コロナで深刻な影響を受けた観光業に対する政府の需要喚起策

Japanese government takes measures to recover demand for the tourism industry severely affected by the Covid-19.

- 旅行代金の割引 + 地域共通クーポン(宿泊以外に利用可能)
   discounts on travel + regional coupons(which is only available for accommodation)
- 高級ホテルにのみ効果が偏るとの指摘 Only luxury hotels benefit from the policy.
  - 旅行代金の割引は宿泊やツアー料金に使途が限定されがちであった Discounts on travel tended to be used only for accommodation.

## 観光事業者を幅広く支援するには、適切な補助制度の設計が必要

Appropriate subsidy system design is needed to support a wide range of tourism businesses.

## Purpose

### 地域共通クーポンが宿泊以外の観光業に与える効果を検証

Verify the effect of the regional common coupons on each sector of the tourism industry other than accommodation.



Transportation



Food



Souvenir

## **Data**

## 道後2017調査(アンケート) Dogo2017-Questionnaire

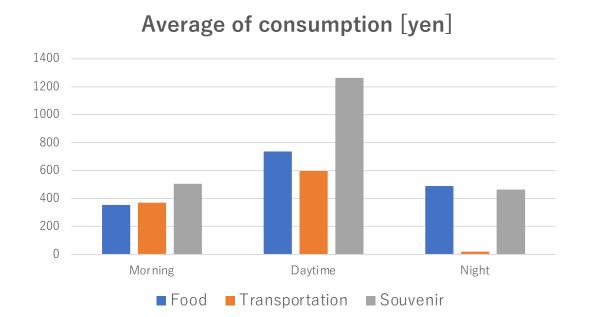
サンプル:道後のホテルに泊まった観光客,有効サンプル197 (松山エリア内で3項目全てで消費行動) Of the tourists stayed at a hotel in Dogo, valid sample 197 -those who record consumption behavior in all three items within the Matsuyama area

旅程に付随する滞在場所・時間帯・消費項目・消費金額のデータを利用 We used data on location, time of day, consumption categories, and amount spent at each of the destinations visited during the trip.

訪問先1/destination							
場所 location	到着時間 arrival	消費1 consumpt	tion1	消費2 consump	tion2	消費3 consumption3	
		項目 category	金額 amount	項目 category	金額 amount	項目 category	金額 amount
道後温泉 Dogo Onsen	16:50	食事 food	2,000	土産 souvenir	3,500	交通 transportation	400

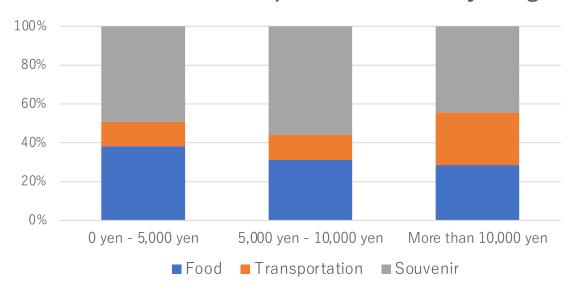
# **Basic Analysis**

## 道後2017調査(アンケート) Dogo2017-Questionnaire



カテゴリ・時間帯ごとに消費金額は大きく違う Consumption varies widely by category and time of day

#### **Differences in Consumption Allocations by Budget**



予算ごとに金額割り振りが変わる
Amount allocations vary by budget

## Method

## 離散連続モデル(Discrete-Continuous model)

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離散量と連続量を同時に選択し、配分するモデル A model that simultaneously selects discrete and continuous quantities and allocates

<b>離散量</b> Discrete カテゴリ category	<b>連続量</b> <b>Continuous</b> 支払い金額 expense	
飲食費 Food or drink	$x_1 \boxminus$	予算制約 $\sum_{k} x_k = E$
交通費 Transportation お土産代	$x_2$ 円 $x_3$ 円 $x_3$ 円	R の中で効用を最大化する ように配分
Souvenir		Distribute expense to maximize utility under the budget constraint

## Method

## MDCEVモデル (MDCEV model)

選択した各カテゴリ(k)に予算 $(x_k)$ を配分する. 効用関数は以下の通り. Allocate a budget to each selected category. The utility function is as follows.

$$U(\mathbf{x}) = \sum_{k} \frac{\gamma_k}{\alpha_k} \exp(\beta z_k + \epsilon_k) \left( \left( \frac{x_k}{\gamma_k} + 1 \right)^{\alpha_k} - 1 \right)$$
 (1)

 $Z_k$  選択肢固有属性や個人属性などの説明変数 (Explanatory variables such as choice-specific attributes and personal attributes)

β パラメータ(parameters)

 $arepsilon_{m k}$  がンベル分布に従う確率項(Probability term following a Gumbel distribution)

α<sub>k</sub> γ<sub>k</sub> 消費の増大に伴う効用の逓減度合いを表すパラメータ(Parameter representing diminishing utility with increasing consumption)

 $\alpha_{\mathbf{k}}$   $\gamma_{\mathbf{k}}$ を同時に推定することは難しいため、 $\alpha_{\mathbf{k}} \to 0$ と固定し、 $\gamma_{\mathbf{k}}$ のみを推定。このとき(1)は Since it is too difficult to estimate  $\alpha_{\mathbf{k}}$   $\gamma_{\mathbf{k}}$  at the same time, we presupposed  $\alpha_{\mathbf{k}} \to 0$  and only estimated  $\gamma_{\mathbf{k}}$ . Then (1) becomes  $U(\mathbf{x}) = \sum_{\mathbf{k}} \gamma_{\mathbf{k}} \exp\left(\beta z_{\mathbf{k}} + \epsilon_{\mathbf{k}}\right) \ln\left(\frac{x_{\mathbf{k}}}{\gamma_{\mathbf{k}}} + 1\right)$ 

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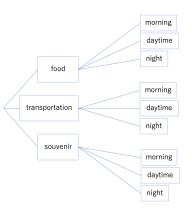
## Method

### 選択構造 Selection Structure

#### 時間帯・支出項目ごとの9つの選択肢

9 alternatives for each time period and category

Morning-Food
Morning-Transportation
Morning-Souvenir
Afternoon-Food
Afternoon-Transportation
Afternoon-Souvenir
Night-Food
Night-Transportation
Night-Souvenir



Why not nested?

→Appendix

# **Detail of Utility Function**

$$V_{morning,food} = eta_{food}$$
 $V_{morning,souvenir} = eta_{souvenir} + eta_{1stvisit,souvenir} imes \delta_{dummy_{1stvisit}}$ 
 $V_{morning,transpotation} = eta_{transportation} + eta_{1stvisit,transportation} imes \delta_{dummy_{1stvisit}}$ 

個人属性 Personal Attributes

 $V_{afternoon,food} = eta_{food} + eta_{afternoon,food}$ 
 $V_{afternoon,souvenir} = eta_{souvenir} + eta_{afternoon,souvenir} + eta_{1stvisit,souvenir} imes \delta_{dummy_{1st}}$ 
 $V_{afternoon,transportation} = eta_{transportation} + eta_{afternoon,transportation} + eta_{1stvisit,transportation} imes \delta_{dummy_{1st}}$ 
 $V_{night,food} = eta_{food} + eta_{night,food} + eta_{night,group} imes \delta_{dummy_{1st}} + eta_{night,group} imes \delta_{dummy_{1st}} + eta_{night,group} imes \delta_{dummy_{3 or more}}$ 
 $V_{night,transportation} = eta_{transportation} + eta_{night,transportation} + eta_{1stvisit,transportation} imes \delta_{dummy_{1st}} + eta_{night,group} imes \delta_{dummy_{3 or more}}$ 
 $V_{night,transportation} = eta_{transportation} + eta_{night,transportation} + eta_{1stvisit,transportation} imes \delta_{dummy_{1st}} + eta_{night,group} imes \delta_{dummy_{3 or more}}$ 

## **Estimation**

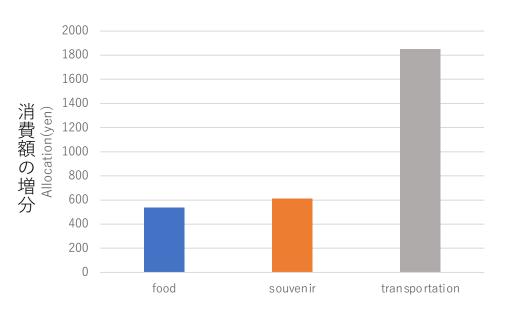
Explanatory Variables	Est.	t-val
food and drink $\gamma$	5.60	3.76**
souvenir $\gamma$	2.76	3.79**
transportation $\gamma$	6.80	4.91**
souvenir ASC	0.713	2.26*
transportation ASC	1.31	4.84**
afternoon food and drink ASC	1.02	4.41**
afternoon souvenir ASC	0.682	2.63*
afternoon transportation ASC	0.216	1.10
night food and drink ASC	0.541	1.81
night souvenir ASC	-1.12	-2.20*
night transportation ASC	-0.365	-1.29
first visit souvenir dummy pram	0.244	0.81
first visit transportation dummy param	0.496	1.93
group of 3 or more night dummy param	-0.982	-3.20**
	* 5% 有意	** 1% 有意

No. observations	197
$\mathcal{L}(0)$	-1121.487
$\mathcal{L}(\hat{eta})$	-918.340
$\rho^2$	0.181
$ar ho^2$	0.194

- 効用の減衰が、お土産代、飲食代、交通費の順に大きい The decay of utility is faster for souvenirs, food and drinks, and transportation costs, in that order
  - → 旅行先では交通費の増加をいとわない!
    While travel, people are willing to pay more for transportation.
- 初めて訪れた人はお土産、交通にお金を使いやすい
   First-time visitors are more likely to spend money on souvenirs and transportation
- 夜はお土産、交通費と比べて飲食費に使いやすい
   In the evening, visitors are more likely to spend money on food and drink compared to other categories

#### 自由に使える3000円クーポンを付与した場合のカテゴリ別消費変化

Change in spending when a ¥3,000 coupon which can be used for every category is given.



#### 仮定

発行したクーポンは全て利用され、 クーポンの使用分だけ予算が増加

#### Assumption

All coupons issued would be used. The budget would increase by the amount of coupons used.

クーポン額の多くはsaturation effectが小さい交通費に使われた

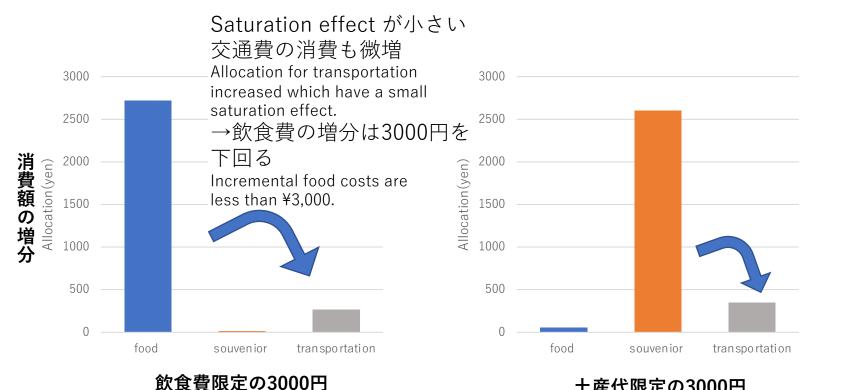
Most of the coupon amount was spent on transportation costs, which have a small saturation effect.

クーポンを配布

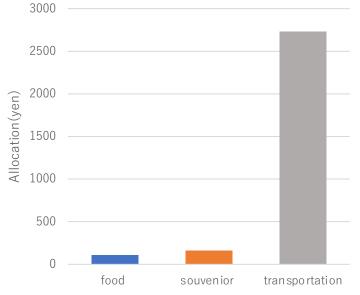
Only for food

#### 各カテゴリ限定の3000円クーポンを付与した場合のカテゴリ別消費額変化

Change in spending by category when a ¥3,000 coupon which is only available for each category is given.



土産代限定の3000円 クーポンを配布 Only for souvenir



交通費限定の3000円 クーポンを配布 Only for transportation

#### 合計5000円分のクーポンをカテゴリ間で平等に配分されるように設計する

Design the allocations so that a total of 5,000-yen worth of coupons are distributed equally among the categories

1000円単位で離散化し、配分のされ方を全通り検証

Discretize in units of 1,000 yen and verify how the allocation is made in all streets.

ex)	

ex <i>)</i>	Subsidy	Allocation
Food and drink	3000円	$x_1$ $oximes$
Souvenir	1000円	x <sub>2</sub> 円
Transportation	1000円	$x_3$ $\boxminus$

平等な配分(=全てに5000/3円ずつ)

からどれほど離れているかを評価

Evaluate how far away from equal distribution (i.e., 5,000/3 yen each for all)

$$Minimize \sum_{k=1}^{3} \left| x_k - \frac{5000}{3} \right|$$

#### 合計5000円分のクーポンをカテゴリ間で平等に配分されるように設計する

(食事, お土産, 交通) = (2000円, 2000円, 1000円)で補助金を出すことが平等配分に最も近くなる (food, souvenir, transportation) = (2000yen, 2000yen, 1000yen) is nearest to equal distribution

 $\gamma_k$ が大きい交通費は、補助金の額が他より低くなる

Transportation costs with larger  $\gamma_k$  have lower subsidies than others

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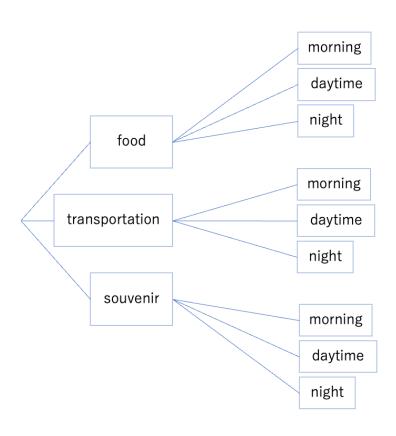
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日本経済新聞. 2021年10月6日,朝刊,GoTo見直し、中小旅館支援を強化 斉藤国交相が示唆.

Matsumura T, Saito H. Designing a tourism stimulus during the Covid-19 pandemic in Japan. RIETI Discussion Paper Series. 21-E-012.

# **Appendix**



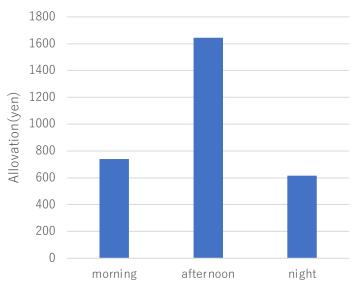
当初はネスト構造を仮定したMDCNEVモデルを 構築したが、スケールパラメーターが1に張り付 いたため、誤差相関がないものとみなしMDCEV モデルとした

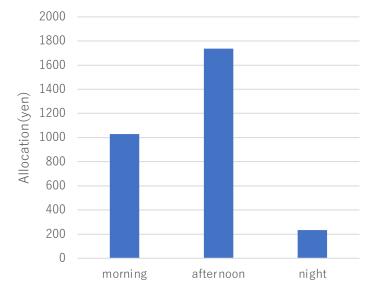
Initially, we constructed MDCNEV model because we presupposed the situation as nested. Since the scale parameter stuck to 1, the MDCEV model was assumed to have no error correlation.

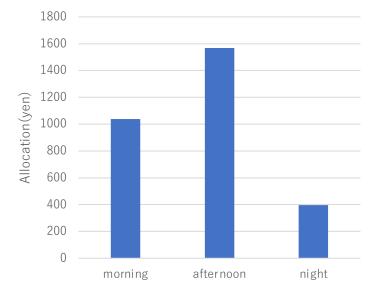
# **Appendix**

#### **Simulation**

# 各カテゴリ限定の3000円クーポンを付与した場合の時間帯別消費変化 Change in spending by time when a ¥3,000 coupon which is only available for each category is given.







飲食費限定の3000円 クーポンを配布 Only for food

土産代限定の3000円 クーポンを配布 Only for souvenir

交通費限定の3000円 クーポンを配布 Only for transportation