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The Cross-Industry Universal Laws of Exponential Booking Curves

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Abstract

In recent years, the e-commerce market has grown with the spread of the internet worldwide every year [1]. Accordingly, in service industries, purchasing products with reservations has become common. With the spread of online reservations, the booking curve, which is the concept of the time series in the cumulative number of reservations and has been used for sales optimization in the airline ticket and hotel industries, has been used in various industries [2–4]. Booking curves in specific industries have been studied, but a universally applicable model across various industries has not been developed [5–8]. Here we show that booking curves can be modeled universally by the exponential decay function, and we also show that the model is valid by using real data from some industries before and after the COVID-19 pandemic, that is, under completely different market conditions. The cross-industry exponential laws of booking curves constitute an important discovery in regard to mathematical laws in the social sciences and can be applied to give leading microeconomic indicators.

Keywords: universal laws, booking curve, exponential decay, economic indicators

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32 1 Methods

- 33 We define a booking curve with a sales deadline that represents the cumulative
 34 number of reservations as follows:

$$35 \quad X(t) \stackrel{\text{def}}{=} \int_t^{\infty} q(s)ds, \quad (1)$$

35 where $q(s)$ is the number of reservations s days before deadlines. Then,
 36 we obtain averaged booking curves (ABCs) for time series $\{X(t; n)|n =$
 37 $1, 2, \dots, N\}$ for N days as follows: $\mathbb{E}[X(t; n)] = \frac{1}{N} \sum_{n=1}^N X(t; n)$. On the
 38 other hand, we model the ABC $\mathbb{E}[X(t)]$ with the following differentiable and
 39 monotonous non-increasing function ϕ :

$$35 \quad \mathbb{E}[X(t)] = A\phi(t; \beta_t), \quad (2)$$

40 where $A(> 0)$ is a capacity constant calculated by the scale of the facility or
 41 the magnitude of demand and $\beta_t > 0$ is an environmental variable given by
 42 the timing of when users make reservations and represents the extent of the
 43 increase (decrease) in the number of reservations over time (t increases). Then,
 44 the function $\phi(t, \beta_t)$ satisfies the following properties:

$$45 \quad \begin{aligned} \phi(0; \beta_0) &= 1, \lim_{t \rightarrow \infty} \phi(t; \beta_t) = 0, \\ \phi(t; \beta_t) &\geq \phi(t+1; \beta_{t+1}) \text{ for } \forall t \geq 0. \end{aligned} \quad (3)$$

45 Here, we set the following assumptions. It is assumed that the environ-
 46 mental variable in the facility is constant; that is, the parameter satisfies
 47 $\beta_t = \beta (= \text{constant})$. This means that the environment surrounding the facil-
 48 ity (for example, the market environment) is constant during the reservation
 49 period, and we call this assumption the uniform demand bath assumption
 50 (UDBA), which is associated with the heat bath in thermodynamics [9].

51 Under this assumption, the ABC of facility F_1 is expressed by Eq. (2).
 52 Considering the state of facility F_1 , going back by δt infinitesimal times, the
 53 number of reservations at time δt is $A\phi(\delta t; \beta) = A'$. Then, we consider a vir-
 54 tual facility, F_2 , with a capacity constant, A' . Since the time δt is infinitesimal,
 55 it is reasonable that the environmental variable in F_2 is equal to β , which is the
 56 same as in F_1 . Here, the number of reservations in the state of going back by
 57 time t_1 for the ABC of F_2 is obtained as $A'\phi(t_1; \beta) = A\phi(\delta t; \beta)\phi(t_1; \beta)$. On
 58 the other hand, this value is equivalent to the number of reservations $A\phi(t_2; \beta)$
 59 when the time is traced back forward to $\delta t + t_1 (\equiv t_2)$ in F_1 . Therefore, we
 60 obtain the following equation:

$$61 \quad \begin{aligned} A\phi(t_2; \beta) &= A\phi(\delta t; \beta)\phi(t_1; \beta) \\ t_2 &= \delta t + t_1. \end{aligned} \quad (4)$$

From Eqs. (3) and (4), function ϕ is uniquely determined as $\phi(t; \beta) = \exp(-\beta t)$ (see supplemental information 4 for an explanation), and we obtain the exponential laws for ABCs as $\mathbb{E}[X(t)] = A \exp(-\beta t)$.

Here, we investigate two issues to verify the universality of our model. First, we study whether some actual data fit into the following approximate expression:

$$\mathbb{E}[X(t; n)] \simeq A \exp(-\beta t). \quad (5)$$

Second, we focus on the reasonability of the UDBA by using data from before and after the COVID-19 pandemic period that have a large fluctuation in the magnitude of demand. For each analysis, we use two years of data from 2019 to 2020 for three facilities in the hotel and car rental industries. The details of the data and facilities are described in supplemental information 6.

2 Results

Fig. 1 illustrates the results based on fitting real data into Eq. (5). First, we investigate the exponential laws of ABCs in terms of the cross-industry, cross-facility, and cross-condition of the market and confirm that real data are almost generally expressed by exponential functions from each perspective. Furthermore, the difference between the environmental variables β across industries or facilities, which are the parameters given by the timing of when people make reservations, results in remarkable differences in products such as resorts and business hotels. Second, regarding the reasonability of the UDBA, Fig. 2 represents the relationship between MSE and CV_τ , each of which is defined as an error based on each fitting exponential function and an indicator of the magnitude of environmental fluctuation. It is confirmed that contents that have large MSE also have large CV_τ ; therefore, this means that the market environment is not stable when ABCs cannot follow exponential laws. As the contraposition of the exponential laws of ABCs under the UDBA, this result also supports the reasonability of our model.

3 Discussions

In the real world, exponential laws have been universally observed in various systems (natural sciences, social sciences, statistical physics, etc.) [9, 12–15]. It is also known that as another universal law, there is a statistical law, such as the super generalized central limit theorem based on the power law [16, 17], which is a ubiquitous characteristic found in such systems. In the present case, considering the generality based on Eqs. (2), (3) and (4), we expect that the exponential laws of ABCs can be *universally* established in industries over those introduced in the Results section.

Furthermore, parameter A , which is derived from the magnitude of demand in the model, is expected to give a leading economic indicator. Fig. 3 illustrates the service industry purchasing managers' index (PMI) in Japan (see

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supplemental information 7 for a description of the PMI) and indicator parameter A based on the exponential model of the ABCs for each month. The PMI is a macroeconomic indicator based on many questionnaires administered to the entire domestic service industry [18, 19], while parameter A is obtained directly from real data for each facility. Fig. 3 confirms that the value of A has changed, especially prior to the emergency declaration period from April 2020 and the tourism campaign period from July of the same year. Meanwhile, indicator A plays a role in demand forecasting for the facility; in addition, the model has more general availability, for example, regarding the area and industry. Hence, we expect the model to serve as an economic model that gives leading microeconomic indicators.

Supplementary information.

4 Proof of leading $\phi(t; \beta)$ into exponential decay functions

We show that a differentiable function ϕ , which satisfies the following Eqs. (6) and (7) under the UDBA in the main paper, is derived uniquely into an exponential decay function.

$$\begin{aligned} \phi(0; \beta) &= 1, \quad \lim_{t \rightarrow \infty} \phi(t; \beta) = 0, \\ \phi(t; \beta) &\geq \phi(t+1; \beta) \text{ for } \forall t \geq 0. \end{aligned} \tag{6}$$

$$\begin{aligned} A\phi(t_2; \beta) &= A\phi(\delta t; \beta)\phi(t_1; \beta) \\ t_2 &= \delta t + t_1 \end{aligned} \tag{7}$$

We obtain the following equation for ϕ independent of A since Eq. (2) is reasonably satisfied for any t_1 or δt under the UDBA:

$$\phi(\delta t + t_1; \beta) = \phi(\delta t; \beta)\phi(t_1; \beta) \text{ for } \forall \delta t, t_1 > 0. \tag{8}$$

Accordingly, we obtain the following identity equation for ϕ for any t and δt with the criterion $\phi(0) = 1$ from Eq. (6).

$$\phi(t + \delta t) - \phi(t) = \phi(t)(\phi(\delta t) - \phi(0)). \tag{9}$$

In addition, the following equation is given by dividing by δt and applying $\lim_{\delta t \rightarrow 0}$.

$$\lim_{\delta t \rightarrow 0} \frac{\phi(t + \delta t) - \phi(t)}{\delta t} = \phi(t) \lim_{\delta t \rightarrow 0} \frac{\phi(\delta t) - \phi(0)}{\delta t}. \tag{10}$$

Facility	Acquisition period	The number of records	Including canceled records
hotel 1		141,828	yes
hotel 2		59,075	no
hotel 3	from Jan. 1. 2019	23,787	no
car rental 1	to Dec. 31. 2020	62,401	yes
car rental 2		334,378	yes
car rental 3		48,472	yes

Table 1 Dataset information used for our verification. We acquired the reservation data of six facilities for two years. One record data represents one piece of reservation information.

123 Then, Eq. (11) results from regarding constant C as $C = \frac{d\phi}{dt} \Big|_{t=0}$:

$$\frac{d\phi}{dt} = C\phi(t). \quad (11)$$

124 Solving this equation, we obtain function ϕ as follows with a certain constant
125 C_0 :

$$\phi(t) = \exp(Ct + C_0).$$

126 Finally, constants C_0 and C , respectively, satisfy $C_0 = 0$ from $\phi(0) = 1$ and
127 $C < 0$ from $\lim_{t \rightarrow \infty} \phi(t; \beta) = 0$ in Eq. (6). Thus, function ϕ is derived
128 uniquely following exponential decay with a certain constant $\beta (= -C) > 0$:

$$\phi(t) = \exp(-\beta t). \quad (12)$$

129 Then, the additional law that is given as Eq. (8) under a certain static environment,
130 such as the UDBA, is also seen in the natural sciences. For example, in
131 thermodynamics, a distribution of energy in a system with constant temperature
132 called a heat bath satisfies a similar addition law. In our case, based on
133 the heat bath concept, we call the criterion a demand bath with static demand
134 environments.

135 5 Dataset for verification

136 Fig. 4 gives an overview of the data used for verification. The original data
137 are published in the repository, except for information that can identify an
138 individual or organization. We define a chartered day in the hotel 1 as having
139 more than 50% group travel customers, which can be identified in the record
140 data. We investigate for periods except for chartered days. All the other facilities
141 do not have chartered days. The number of records have all of the data
142 including chartered days.

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No.	Period	Remark for demand
2019	Jan.1 - Dec.31(2019)	normal demand
2020	Jan.1 - Dec.31(2020)	COVID-19 pandemic
2020-1	Jan.1 - Mar.31(2020)	the first COVID-19 epidemic
2020-2	Apr.1 - Jul.21(2020)	emergency declaration
2020-3	Jul.22 - Dec.31(2020)	tourism campaign

Table 2 The information of periods divided in terms of the market conditions for our verification.

143 6 Periods divided by market conditions

144 Tab. 1 (This is the same as the table in Fig. 1. b in the main paper.) shows
 145 how we divide the periods to validate the exponential laws of averaged booking
 146 curves (ABCs). In 2019 and 2020, the market condition changed significantly
 147 due to the influence of the COVID-19 pandemic. Thus, we compare each
 148 of them based on annual data. In addition, in 2020, the Japanese govern-
 149 ment implemented some strategies related to the movement of people, that
 150 is, directly related to the demand environment for services such as hotels and
 151 car rentals. Therefore, we divide the periods based on these strategies. In the
 152 period from Jan. 1 to Mar. 31, COVID-19 spread in Japan for the first time;
 153 accordingly, the number of people from overseas gradually decreased. In the
 154 period from Apr. 1 to Jul. 21, the Japanese government declared a state of
 155 emergency due to COVID-19 and strongly requested citizens to self-restrict
 156 going out. In the period from Jul. 22 to Dec. 31, the government implemented
 157 the tourism campaign called "Go to Travel", which was a deep discount strat-
 158 egy for domestic travel products; therefore, people had more opportunities to
 159 go out.

160 7 PMI: Purchasing managers' index

161 The purchasing managers' index (PMI) is an economic indicator derived from
 162 a monthly survey of private companies. It is calculated based on the answers
 163 provided by companies' purchasing managers or executives regarding, for
 164 example, their firms' production, employment, inventories, order backlogs, new
 165 export orders and imports of materials and supplies. In addition, a PMI value
 166 exceeding 50 indicates better business conditions than in the previous month.
 167 Therefore, the PMI is often used as a leading economic indicator, and some
 168 economic models have been studied by using the PMI. In this study, we pro-
 169 pose that an exponential function model of booking curves using reservation
 170 data in service industries can be applied as an economic model that gives a
 171 more microscopic and real data-driven economic leading indicator along with
 172 the PMI. The main paper introduces that the value of A estimated using the
 173 exponential function model can be used as a leading economic indicator (or as
 174 one of the samples that composes it). Note that our model has three features.
 175 First, A can be calculated directly from the data; second, A can be calculated

176 at the beginning of the month; and third, A can be calculated in units smaller
177 than the PMI, such as facilities and areas.

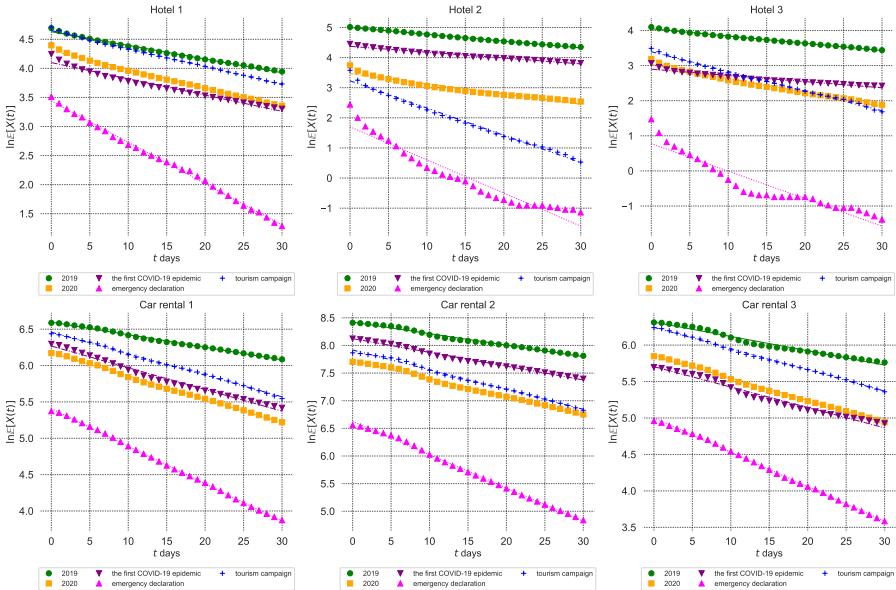
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Industry	Facility region	Area type	Characteristics
a	hotel 1	Tochigi	regional city
	hotel 2	anonymous	city
	hotel 3	anonymous	city
	car rental 1	island Ishigaki	resort
	car rental 2	island Naha	resort
	car rental 3	island Miyako	resort
Period	Remark for demand		
2019	Jan.1 - Dec.31(2019)	normal demand	
2020	Jan.1 - Dec.31(2020)	COVID-19 pandemic	
2020-1	Jan.1 - Mar.31(2020)	the first COVID-19 epidemic	
2020-2	Apr.1 - Jul.21(2020)	emergency declaration	
2020-3	Jul.22 - Dec.31(2020)	tourism campaign	

Fig. 1 The real logarithmic ABCs data with for two years, two industries and six facilities. This figure illustrates how real data universally fit into the exponential laws with Eq. (5) in terms of three perspectives: the cross-industry, cross-facility and cross-condition of the market. We use reservation record data on six facilities in the hotel and rental car industries for validation with Eq. (5), with a fitting range of $0 \leq t \leq 30$. First, we investigate each facility in 2019 (green lines), where the market environment was relatively stable; that is, the UDBA would be satisfied. It is almost confirmed that the data follow exponentially in five facilities except car rental 3, which means that the laws are common across industries. Second, we focus on the differences in characteristics such as business and resort hotels in 2019. Although hotels 1, 2 and 3 are within the same industry, the attenuation coefficients of reservations (slopes; derived from β) are different; therefore, parameter β is significantly consistent with the nature of the facility and product since it changes due to the difference in when users make reservations. Third, we compare three periods of different demand in 2020 (details in supplemental information 6 [10, 11]), which do not necessarily satisfy the UDBA (purple, blue and pink lines). Nevertheless, in each facility, the intercept (derived from capacity constants) based on the magnitude of demand and the slopes (environment variables) differ under each period, and we find the exponential laws in some facilities and periods.

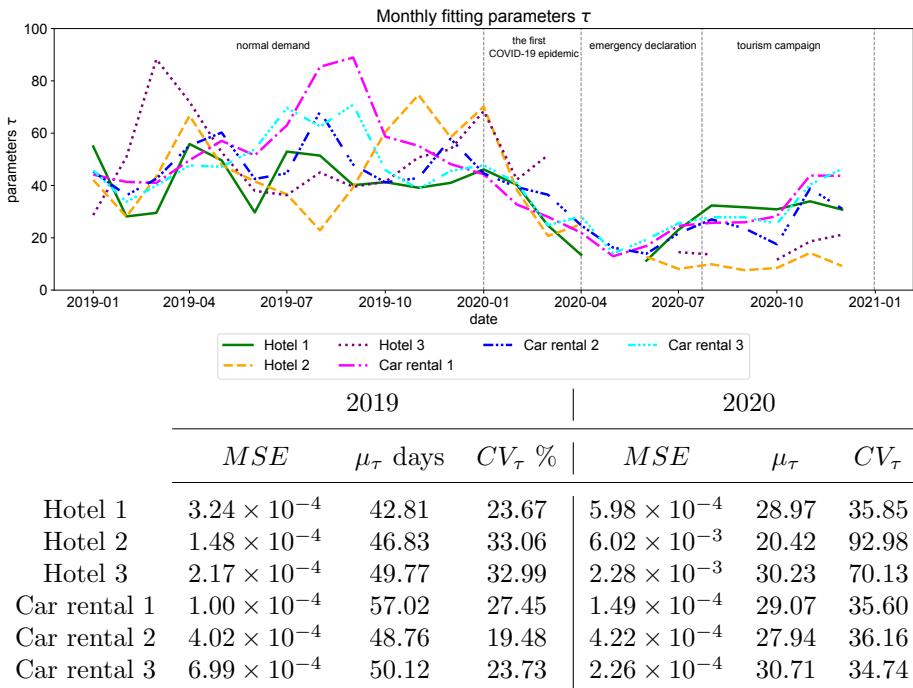


Fig. 2 The relation between the fluctuation of the environment parameter τ derived from monthly data and the error between logarithmic ABCs obtained from annual year and their regression lines. This figure illustrates how the exponential laws of ABCs relate to the environment around facilities quantitatively by using data from 2019 to 2020. The line graph shows the transition of the time constants of exponential function τ_{ij} , which is given as $\tau = 1/\beta$ [12] in Eq. (5) in year and month ij . The table shows the three indicators given below for 2019 and 2020. First, MSE is the mean square error that represents how ABCs deviate from exponential functions as seen in green and yellow lines in

Fig. 1. It is defined as follows: $MSE(N; t^*) = \frac{1}{t^*} \int_0^{t^*} |\ln(\mathbb{E}[X(s; n)]) - (\ln A - s/\tau)|^2 ds$, where N is the number of days in each period, that is, the number of booking curves $X(t)$ to be averaged; additionally, A and τ are the fitting parameters obtained by Eq. (5) with $\tau = 1/\beta$. The constant t^* , which is given as $t^* = 30$ in our observation, gives integration intervals. Second, μ_τ is the annual average of τ_{ij} obtained from the ABCs in year and month ij ; thus, it is given by $\mu_{\tau_i} = \mathbb{E}_j[\tau_{ij}]$. Third, CV_τ is the annual coefficient of variation of τ_{ij} [12]; thus, it is defined as follows: $CV_{\tau_i} = \frac{\sigma_{\tau_i}}{\mu_{\tau_i}} \times 100$, where $\sigma_{\tau_i} = \sqrt{\mathbb{E}_j[(\tau_{ij} - \mu_{\tau_i})^2]}$.

The table confirms that the MSE errors in hotels 2 and 3 in 2020 have a larger order by 10^{-3} than other facilities and than each previous year. Furthermore, CV_τ , which indicates the magnitude of environmental fluctuations, is large. Thus, the deviation from the exponential laws results from the instability of the environment, for example, the COVID-19 pandemic.

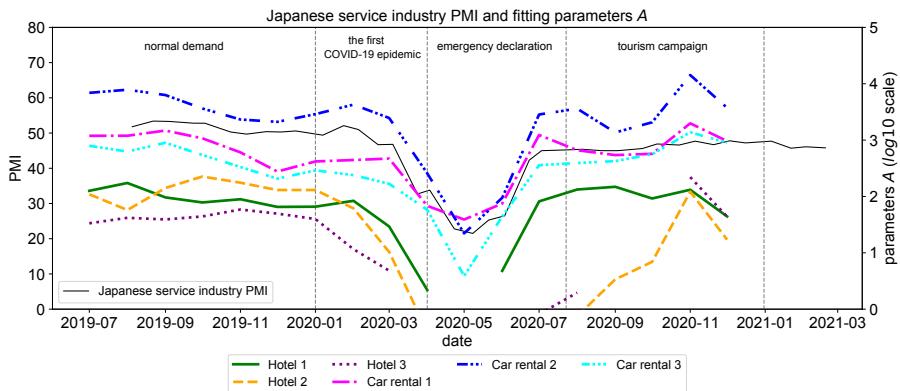


Fig. 3 The relation between Japanese service industry PMI and the estimated demand indicator A in each facility at the beginning of a month. We discuss the capability of the exponential laws of ABCs as an economic model in this figure. The graph illustrates Markit's Japanese service industry PMI, which is a leading economic indicator published on markiteconomics.com, and the transition of the monthly fitting parameter A_{ij} in year and month ij , where $30 \leq t \leq 60$ in Eq. (5). Then, since parameter A is given at the beginning of the month, it can represent an indicator of the economics surrounding a facility. In fact, it is confirmed that the fitting parameters A similarly decreased or increased prior to the emergency declaration period from April 2020 and the tourism campaign period from July of the same year.