

CHAPTER 6

Modelling Internet Diffusion Across Tourism Sectors

Miriam Scaglione and Jamie Murphy

INTRODUCTION

Everett M. Rogers' (1962) book, *Diffusion of Innovations* (DOI), the most cited work in innovation research, discusses innovation adoption and implementation at the individual and organisational level. Rogers (2003, p. 12) defines an innovation as "an idea, practice, or object that is perceived as new by an individual or other unit of adoption". Rogers' book proposes that diffusion follows a normal distribution and therefore, fixed percentages for his five—innovators, early adopters, early majority, late majority and laggards—adopter categories. These fixed percentages, however, rarely hold across countries and innovations.

This chapter draws on a pedagogical explanation of Bass modelling of Swiss accommodation in (Scaglione, forthcoming), and a first draft of this paper presented at the Annual Conference of International Association of Scientific Experts in Tourism—AIEST 2012 held at Khon Kaen (Thailand).

- A1 M. Scaglione (✉)
 A2 Applied University of Western Switzerland, Sierre, Switzerland
 A3 e-mail: miriam.scaglione@hevs.ch
 A4 J. Murphy
 A5 University of Eastern Finland, Kuopio, Finland

Similar to the diffusion of many innovations, internet adoption follows an abnormal distribution due to complex and diverse economic, political, cultural, infrastructural and geographical factors (Andrés et al. 2010; see Kim 2011). Rather than such macro-factors, this study adds communication channels (Bass 1969) and heterogeneous adopter categories (Bemmaor 1994) to improve modelling innovation adoption.

Rogers proposes that at the beginning of the adoption process a few innovators influence others using intrapersonal communication such as word-of-mouth. But if the role of communication channels, such as word-of-mouth, have the same importance along with the adoption process is an open question in the literature (Easingwood et al. 1983).

Incorporating adopter category differences, abnormal diffusion rates and differing adoption drivers improves the diffusion calculations. These calculation results are critical for industry to predict market size and sales of an innovation. This study takes two small steps in modelling the organisational adoption of innovations, adding a threshold due to a critical mass of adopters and heterogeneity adoption propensities. These two factors help compare innovations across countries and tourism sectors.

This modelling story begins by reviewing two improvements to Roger's diffusion model. Bass' (1969) model predicts specific—industry, product and country—new product adoption rates and market sizes. Bemmaor (1994) complements Bass' innovation-specific adoption model by incorporating heterogeneous rather than homogenous adoption factors at the actor level. The results of these two models, across 13 tourism sector datasets, add to a growing body of innovation-specific adoption research and key factors related to internet diffusion in the tourism sector. Based on these two analyses, this paper proposes future research streams related to modelling the DOI.

LITERATURE REVIEW

The tourism sector has been at the forefront of innovation adoption and implementation at least since the dawn of airline reservation systems. A chance meeting of IBM and American Airlines executives in 1953 led to the development of Sabre, an automated system to help manage the link between airline inventory and customer reservations (Sabre, n.d.).

Since that moment, information and communication technologies (ICTs) have been a key tool for marketing and distribution in tourism (O'Connor 1999). ICTs offer opportunities information dissemination

(24*7*365) on goods and services, pricing strategies and promotions (last-minute, location-based offers, etc.), as well as broaden selling opportunities for tourism suppliers including getting customer feedback from social media sites, no matter the size of the company (O'Connor and Murphy 2004). Therefore, tourism organisations and individuals have been at the forefront of using ICTs such as websites, email and social media.

Understanding and predicting tourism ICT adoption rates, though, remains an ongoing challenge.

Innovation and Imitation

Bass (1969) expanded Rogers' (1962) adoption model by quantifying two factors that drive individual and organisational adoption, *innovation* (p) driven by external channels such as mass communication and *imitation* (q) driven by intrapersonal communication such as word-of-mouth. Bass' model (1969) calculates innovation-specific and abnormally shaped distribution and adoption curves. This non-cumulative distribution curve has three important values that set the adopter categories. The inflexion point $T1$ separates the early adopters and early majority categories. Next, the peak of absolute adoption T demarcates the early majority and late majority categories. Finally, the second inflexion point $T2$ delineates the late majority and laggards.

Equation (6.1) below, a modified exponential and logistic function (Meade and Islam 2006), shows the Bass function (1969) for adopting an innovation. $N(t)$ is the cumulative number of adopters at time t , m is the total market adoption. Parameters p and q , the coefficients of innovation and imitation, respectively, fuel the adoption.

$$\frac{dN(t)}{dt} = \underbrace{\frac{p(m-N(t))}{m}}_{\text{adoption due to external influence or independent adoption}} + \underbrace{\frac{q}{m}(m-N(t))}_{\text{adoption due to internal influence or internal adoption}} \quad (6.1)$$

The coefficients p and q , respectively, suggest the propensity to adopt driven by external information and interpersonal communication channels (Mahajan et al. 1990). With pure innovation, $q=0$ and $p>0$ and $p=0$ and $q>0$ with pure imitation.

Apart from these extreme cases, the sum and ratio of q and p delineate the diffusion curves. Their sum controls the scale of the S-shaped

85 cumulative and non-cumulative adoption curve (respectively Figs. 6.1
 86 and 6.2), and the bell-shaped non-cumulative adoption curve. The q/p
 87 ratio indicates the shape of the cumulative and non-cumulative adop-
 88 tion curves. The bigger the q/p ratio, the more prominent the S-shape
 89 and slower penetration rates are as shown in Fig. 6.1 (Bemmaor and Lee
 90 2002; Meade and Islam 2006). Comparing the p and q parameters helps
 91 examine adoption across countries and the adoption of different innova-
 92 tions by the same population.

93 Van den Bulte and Stremersch's (2004) study of individual-level q/p
 94 ratios for 52 consumer durables across 28 countries found, among other
 95 aspects, a positive relationship between the q/p ratio and income ine-
 96 quality. The bigger a country's income inequality, the more that internal
 97 rather than external influences—respectively imitation and innovation—
 98 drove adoption. These authors argue that the adoption of the same
 99 product in different countries will show different q/p ratios.

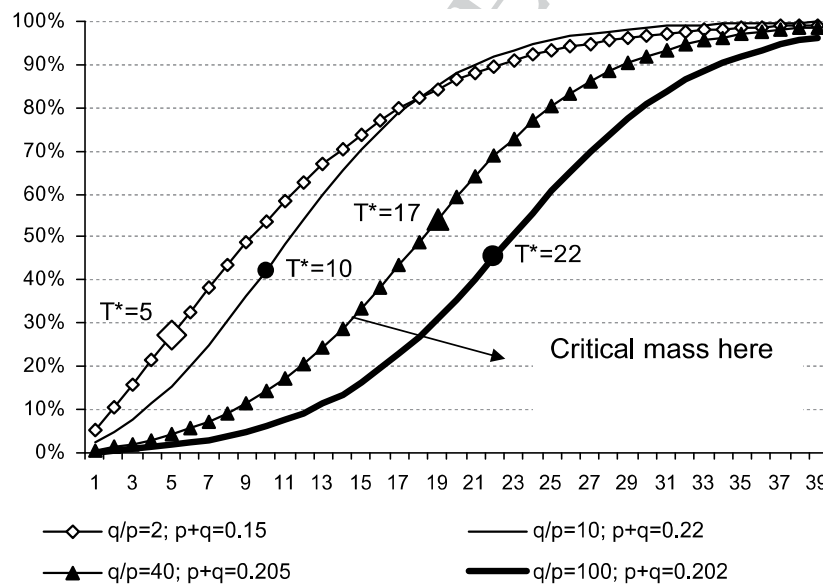


Fig. 6.1 Cumulative bass distribution models with different q/p ratios; T^* is the peak rate of adoption, namely the inflexion point of the cumulative distribution (*Source* Authors' computation)

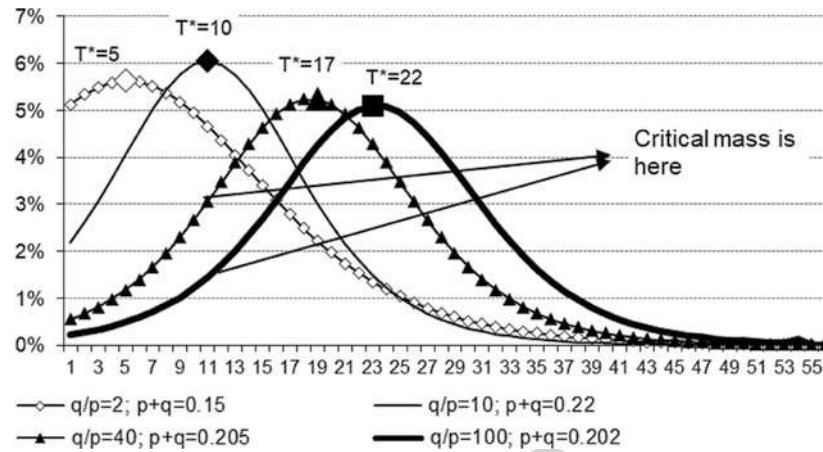


Fig. 6.2 Non-cumulative bass distribution models with different q/p ratios; T^* is the peak rate of adoption, namely the inflexion point of the cumulative distribution (Source Authors' computation)

Homogenous and Heterogeneous Adoption

The Bass model as in Eq. (6.1) assumes external communication—e.g. price, advertising and promotion—and intrapersonal communication such as word-of-mouth drive adoption. An extension of Bass' distribution curves helps bridge Rogers' (2003) five adopter categories into the Bass/Rogers or BR model (Mahajan et al. 1990). Rather than using the timing of adoption, the BR categories assume a homogenous likelihood to adopt across the five adopter categories (Bemmaor 1994, p. 203).

Bemmaor (1994) proposes a general diffusion model with a heterogeneous propensity for the first-adoption (purchase) in order to address these limitations. At the individual adopter level, the propensity to adopt an innovation at time t follows a shifted Gompertz distribution (SG) whereby early adopters take up the innovation in a more random fashion than latter adopters. Under this assumption, Bemmaor and Lee (2002) propose the Gamma/shifted Gompertz (G/SG) model of Eq. (6.2) below,

$$F(t) = \left[1 - e^{-bt}\right] / \left[1 + \beta e^{bt}\right]^{\alpha} \quad (6.2)$$

AQ3

The β parameter of Eq. (6.2) is its scale parameter; b is the scale parameter of the shifted Gompertz prior distribution of the individual propensity model. Bass model parameters, $b = p + q$ and $\beta = q/p$, can replace b and β . For fixed values of b and β , the shape of the Eq. (6.2) gamma distribution function depends on the parameter α , which measures adopter population heterogeneity. The coefficient of variation of a Gamma function is $\alpha^{-1/2}$ (standard deviation = $b\alpha^{1/2}$ over the mean = $b\alpha$).

Homogeneity has a direct relationship with α , and when α tends towards ∞ , the population is homogenous. The propensity to adopt is roughly the same across potential adopters. When $\alpha \approx 1$, the model is the Bass model (Eq. 6.1). When $\alpha \approx 0$, potential adopters' acceptance rates differ across strata population. Bemmaor (1994, p. 220) suggests that the propensity to adopt differs across the BR adopter categories.

129 *Critical Thresholds and Exogenous Variables*

In addition to ignoring heterogeneous adoption propensities among the population, the Bass model as expressed by Eq. (6.1) ignores the critical threshold necessary to adopt (Van den Bulte and Stremersch 2004). This threshold reflects the requisite critical mass of adopters. Analysis of the q/p ratio gives insights on critical mass' importance for comparing diffusion across countries or comparing innovations across the same adopter population. For example, a qualitative study of German banks showed that relative to non-interactive innovations, the diffusion of interactive innovations (i.e. Electronic Funds Transfers and home banking for private customers) was slow until reaching a critical mass of adopters (Mahler and Rogers 1999). If this finding is the case, then the S-Shape for the non-interactive innovations should be less pronounced and the q/p ratio smaller than for interactive innovations as shown in Fig. 6.1.

The critical threshold can also depend on exogenous, external or environmental factors. At the organisational or aggregate-level, perceived innovation profitability (Jensen 1982) and risk-averse behaviour (Oren and Schwartz 1988) may accelerate or delay adopting an innovation. At the individual level, Chatterjee and Eliashberg (1990) used a micro-modelling approach to show that initial adopter value perceptions, aversion and perceived information reliability were both heterogeneity sources and threshold factors (see Van den Bulte and Stremersch 2004, p. 534).

AQ4

AQ5

The literature review shows that the Bass model (Eq. 6.1) cannot capture heterogeneous adoption dispositions across a given population of adopters or a given sector. Nevertheless, the q/p ratio helps capture critical mass differences across sectors. These two main facts, namely heterogeneous adoption and critical mass, with an innovation's final penetration rate allow comparing adoption processes of an innovation across sectors.

This research compares internet diffusion across tourism sectors/countries, firstly by analysing the Bass parameters (Eq. 6.1), particularly the q/p ratio, in order to grasp the requisite critical mass for each sector/country and the different adoption curve shapes. Next, the analysis compares heterogeneity across these sectors/countries via the G/SG model.

DATA AND METHODOLOGY

Table 6.1 shows characteristics of the 13 datasets such as the sample and gathering the time of first adoption. Ten studies used the internet Archives' Wayback Machine (archive.org) to gather the website age (Murphy et al. 2007). The other three studies, all related to Swiss accommodation, used a customised softbot, or software robot (Steiner 1999). The softbot gathered the initial domain name registration date, registrant name and location for all Swiss (.ch) domain names registered from December 1995 to February 2004, which contained the keywords "auberge, gasthaus or hotel".

The datasets range in size from 96 to 2467 organisations and range geographically across the globe, Europe and four countries—Austria, Germany, Malaysia and Switzerland. The six tourism sectors in the datasets comprise cable cars/ski lifts, destination management organisations (DMOs), one- to five-star hotels, guesthouses, restaurants and tour operators. The analysis for both the Bass and G/SG model parameters draw on the SAS V9 Nonlinear Least Square Proc Model procedure (SAS Institute Inc. 2011).



Table 6.1 Study sample

Acronym	Sector	Age source	Market size (base year)	Source	Website age/n (URL)	Observed penetration rate respect to 1)
DMO CH	Swiss DMOs	WM	155 (2005)	myswitzerland.ch	149/155	100%
DMO AU	Austrian DMOs	WM	na	Klimek et al. (2011)	96/96	Na
DMO DE	German DMOs	WM	na	Klimek et al. (2011)	182/188	Na
Restaurant CH	Swiss restaurants	WM	18,867 (2005)**	Swisscom Directories	1573/1858	10%
Cable car CH	Swiss cable car companies	WM	370 (2010)	Seilbahnen Schweiz (seilbahnen.org)	190/190	51%
Hotel Chain	International hotel chains	WM	325 (2010)	hotelsmag.com (July 2006)	267/276	85%
European tour operators	European tour operators	WM	na	etooa.org , european-travel-market.com	117/121	na
Malaysian hotels	Malaysian hotels	WM	530	Hachim et al. (2012)	305/315	60%
No affiliated hotel CH	Unaffiliated Swiss hotels	Softbot	1133 (2003)	Swisscom Directories	780/780	69%
Hotelleriesuisse CH	Affiliated Swiss hotels	Softbot	2122 (2003)*	Scaglione et al. (2010)	1677/1733	82%
Hotel CH	Swiss hotels	WM	3255 (2003)*	Scaglione et al. (2010)	2467/2513	77%
Guesthouse CH	Swiss guesthouses	Softbot	3463 (2003)*	Scaglione et al. (2004)	2250/2269	65%
Travel agencies CH	Swiss travel agencies	WM	272 (2008)	Schweizerischer Reisebüro-verband (www.srv.ch)	244/232	90%

Source: Recensement des entreprises 2005, Office fédéral de la statistique (OFS); Calculated from statistique de l'hébergement touristique (HESTAS, OFS)

AQ6

AQ7

RESULTS

184

185 Table 6.2 shows the model estimation results, such as the coefficients of
186 innovation and imitation.

187 Figures 6.4, 6.5, 6.6, 6.7, 6.8, 6.9, 6.10, 6.11, 6.12, 6.13, 6.14,
188 6.15, and 6.16 in the Appendix show the following cumulative series
189 for each sector: *observed*, i.e. the actual value, *estimated* values namely
190 one-step-ahead estimation and *forecasted* values for the last 24 months
191 of the sample using Bass model for each of the sector under study.
192 Table 6.3 in the Appendix shows the goodness of fit measures for the
193 Bass model estimation and forecasts observations.

194 Three Swiss sectors—cable cars, guesthouses and DMOs—have a
195 logistic distribution. Only imitation, influence via intrapersonal commu-
196 nication, drives their internet adoption. Both innovation and imitation,
197 respectively, external and intrapersonal communication, drive internet
198 adoption in the ten other sectors with a 10.42 q/p median. The q/p ratio,
199 intrapersonal communication divided by external communication, shows
200 each channel's relative strength in the adoption process.

201 A high q/p ratio shows that imitation relative to innovation drives
202 adoption. Similar to the three Swiss sectors with a logistic distribution,
203 three additional Swiss sectors have the highest q/p ratios, ranging from
204 26 to 35. Using two measures of age, domain and website, these six
205 Swiss tourism sectors adopt due to imitation more so than innovation.
206 There seems a general parsimony of Swiss sectors, sensible to the critical
207 mass of adopters and a wait and see behaviour.

208 At the innovation end of the communication channel ratio, Austrian
209 DMOs are just under three with a $q/p=2.94$. Hotels chains and
210 European TOs also had q/p ratios under six. More so than the other
211 ten sectors, these sectors adopted due to innovation, that is, persuasion
212 via mass communication. The median group—Malaysian hotels, Swiss
213 restaurants, Swiss travel agents and German DMOs—had imitation/
214 innovation ratios towards 10. The effects of channel communication
215 accelerated the adoption process. As an example, Table 6.2 shows that
216 Austrian DMOs had the lowest q/p (2.94) and the time to generate the
217 early majority group was only 21 months against 48 months for the late
218 majority. In contrast, Swiss hotels with the highest (35.03) ratio, took
219 42 months to generate the early majority against only 17 months for the
220 late majority.



Table 6.2 Estimated Bass and G/SG α parameter with its standard deviations in (parentheses)

Type of model	Drivers	Final market size	Final market size respect to 1)	Initial & final obs.' dates	Bass model (Eq. 6.1)				G/SG			
					T1	Peak	T2	p	q	q/p	Alpha	
					Inflection point	Maximum point	Inflection point	Innovation	Imitation	q/p		
DMO CH	Bass $p=0$ (Logistic)	150.1 (18.72)	96.77%	Dec-96–Feb-05	Jul-99	Jul-99	Jan-01	0 (na)	0.0661 (0.0113)		0.230 (0.023)	
DMO AU	Bass	96.78 (17.02)	na	Dec-96–July-10	Sep-99	Sep-99	Jun-03	0.0079 (0.0021)	0.0232 (0.0171)	2.94 (1.34E7)	1.97E8 (1.34E7)	
DMO DE	Bass	181.63 (21.96)	na	Nov-96–Oct-10	Dec-00	Dec-00	May-04	0.0050 (0.0022)	0.0503 (0.0153)	10.14 (0.09)	0.81 (0.09)	
Restaurant CH	Bass	1647.35 (70.04)	8.73%	Nov-96–Jun-08	Apr-99	Nov-01	Jun-04	0.0035 (0.0011)	0.0392 (0.0067)	11.19 (3.37)	171.6 (3.37)	
Cable car CH	Bass $p=0$ (Logistic)	194.5 (1.30)	52.70%	Dec-95–Feb-05	Jan-00	Jan-01	Jan-02	0	0.0875 (0.0018)		0.61 (0.07)	
Hôtel chain	Bass	269.2 (38.56)	82.77%	Dec-96–Nov-06	Feb-99	Jan-99	Apr-01	0.0099 (0.0101)	0.0376 (0.0165)	3.80 (0.056)	0.51 (0.056)	
European tour operators	Bass	119.18 (9.39)	na	Oct-96–Feb-06	Aug-99	Aug-99	Nov-01	0.0076 (0.0049)	0.0399 (0.0205)	5.28 (0.98)	0.98 (0.28)	
Malaysian hotel	Bass	316.9 (12.8)	59.81%	Nov-96–June-08	Oct-98	Jul-01	Apr-04	0.0037 (0.0015)	0.0364 (0.0093)	9.85 (0.35)	0.35 (0.02)	
No affiliated hotel CH	Bass	796.4 (12.0)	70.26%	Dec-95–Feb-04	Sep-98	Dec-99	Apr-01	0.0029 (0.0014)	0.0799 (0.0054)	27.39 (0.74)	0.74 (0.04)	
Hôteleriesuisse CH	Bass	1716.6 (22.0)	81.93%	Dec-95–Feb-04	Feb-98	Jul-99	Jan-01	0.0028 (0.0007)	0.0718 (0.0035)	25.56 (1.16)	1.16 (0.07)	
Hôtel CH	Bass	2525.5 (25.2)	77.57%	Dec-95–Feb-04	Jan-97	Jul-99	Nov-01	0.0022 (0.0007)	0.0775 (0.0030)	35.03 (1.27)	1.27 (0.08)	
Guesthouses CH	Bass $p=0$ (Logistic)	3131.0 (235.5)	90.41%	Dec-95–Dec-03	Jul-01	Oct-03	May-06 ^a	0	0.0521 (0.0022)		No convergence	
Travel agencies CH	Bass	235.13 (39.04)	98.72%	Nov-96–May-06	Sep-00	Apr-02	Dec-03	0.0063 (0.0033)	0.0676 (0.0563)	10.70 (0.60)	0.60 (0.08)	

^aSource: Authors' calculations

Market Penetration

The median forecasted internet penetration rate (m /total population) for the 10 sectors possible to fix the total market size was 80%. The penetration rate ranges from 9% for Swiss restaurants to over 90% for Swiss travel agents, Swiss guesthouses and Swiss DMOs.

Below the median are Swiss cable cars, Malaysian hotels and non-affiliated Swiss hotels. These sectors tend to comprise small organisations and have low levels of internationalisation, which can help explain their low level of final penetration. The low penetration of the accommodation sector could also be a *leapfrog effect*, whereby late adopters may forego traditional websites in favour of a new technology such as social networks (Hashim et al. 2012).

Close to the median, Swiss hotels (overall and affiliate) and international chain hotels, show a consistent accommodation sector behaviour. In the highest level are Swiss guesthouses, DMOs and travel operators. The Swiss government's push towards internet adoption could help explain the strong DMO adoption (Scaglione and Schegg 2015). Swiss travel agencies seem to follow the general trend of e-tourism. The increasing role of the internet in holiday planning by consumers and businesses may have spurred travel agencies to follow this tendency in order to meet their market.

Homogeneity and Heterogeneity

The α values in the G/SG models reflect homogeneity, with Austrian DMOs followed by Swiss restaurants at the highest levels. Swiss hotels, affiliated and unaffiliated, and European tour operators are $\alpha \approx 1$. The sectors, listed by increasing are German DMOs, Swiss unaffiliated hotels, cable cars, travel agents, and chain hotels, Malaysian hotels and finally Swiss DMOs. Only sectors showing α near 1 follow a Bass model, thus having a similar propensity of first adoption across companies and also among BR categories.

The other Swiss accommodation sector, non-affiliated hotels, shows a greater heterogeneity than overall Swiss hotels and affiliated hotels, in line with the assumption that unaffiliated hotels are smaller than affiliated hotels. The high α value reflects differences within these sectors.

Sectors showing the greatest homogeneity are Swiss restaurants and Austrian DMOs. The underlying reason for Austrian DMOs are

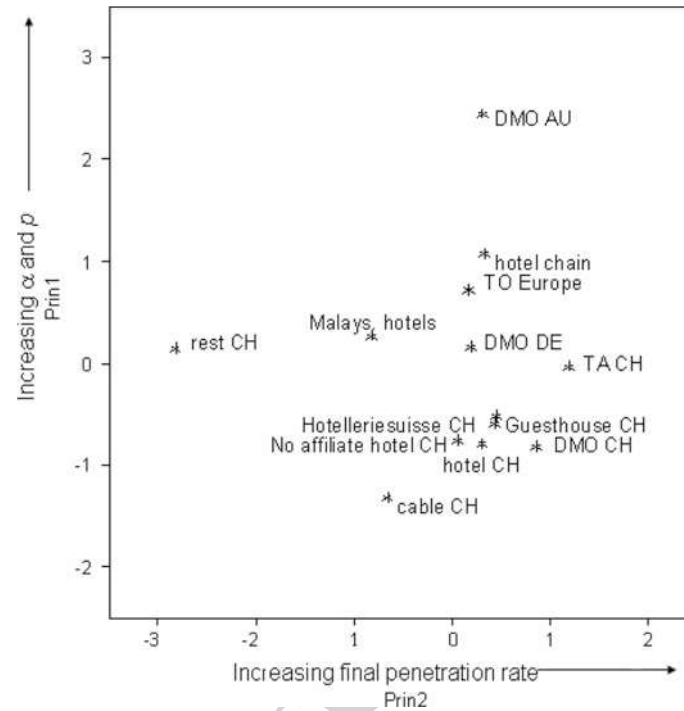


Fig. 6.3 Principal components analysis of Bass parameters, α level of heterogeneity and final penetration rate across tourism sectors/countries (*Source* Authors' computation)

257 beyond the understanding of the authors but the homogeneity of Swiss
 258 restaurants could be explained by the lowest internet penetration level
 259 of 9%. A careful study of the restaurant data shows that its homogeneity
 260 could be due to location. Swiss regions with high tourism intensity like
 261 the cantons of Bern, Vaud and Valais, have more restaurants in comparison
 262 to all others cantons.

263 To shed light on the mixed role of the four values— p , q , penetration
 264 rate and α —a principal components analysis using the Proc Princomp of
 265 SAS V9.2 yields the graph in Fig. 6.3. The two first components explain
 266 a total variation of 77%. Prin1, on the y-axis, shows a 0.59 positive correlation
 267 with p , a 0.52 positive correlation with α and a 0.62 negative

correlation with q . This first component seems to capture the diffusion models parameters. Prin2 on the x -axis reflects the final penetration rate, showing a positive 0.96 correlation with predicted market penetration. The Swiss sectors clustered in the graph, excluding Swiss restaurants, suggest a possible influence of culture/national factors. Another clustering of international hotels chain and European tour operators suggests a possible influence of internalisation and company size.

DISCUSSION AND CONCLUSION

The analysis shows both the importance of innovation and imitation parameters, homogeneity, critical mass and final market size, and that the 13 tourism sectors varied. Almost all Swiss sectors reflect a high sensibility to critical mass. Possible organisational and cultural factors include resistance to innovation and aversion to risk. Unlike the other Swiss sectors, Swiss DMOs show high heterogeneity. This diverseness aligns with the country's different linguistics/cultural factors and organisational differences such as the size—local, regional or national—budget and intensity of the tourism they represent.

The two international datasets illustrate the importance of testing homogeneity. With near homogenous adoption across categories, European tour operators follow the Bass model of adoption. International hotels chains, however, illustrate the importance of modelling heterogeneity.

As future research streams, quantitative and qualitative studies should continue ferreting out the drivers and barriers to both technology adoption and organisational implementation of that same technology. Qualitative findings can lead to improvements in modelling the diffusion of innovations. The empirical research will enlighten comprehension of the tourism sector and improve diffusion forecasts, particularly the take-off phase that many models ignore.

Limitations of this research are mostly related to the gathering of data process as it uses two different methods, one of them (softbot) only collects the “.ch” domain. Hence the comparison across Swiss accommodations sectors deserves a cautious attitude. Another limitation is that the only parameter reported of G/SG model is heterogeneity (α). This report does not fully address the latter model and future research should make a deeper results analysis.

APPENDIX

Cumulative series, Observed, estimated (one-step ahead) and fore-casted for the last 24 years using Bass model for all sectors under study.

See Figs. 6.4, 6.5, 6.6, 6.7, 6.8, 6.9, 6.10, 6.11, 6.12, 6.13, 6.14, 6.15, and 6.16.

Residual and forecasting errors statistics.

See Table 6.3.

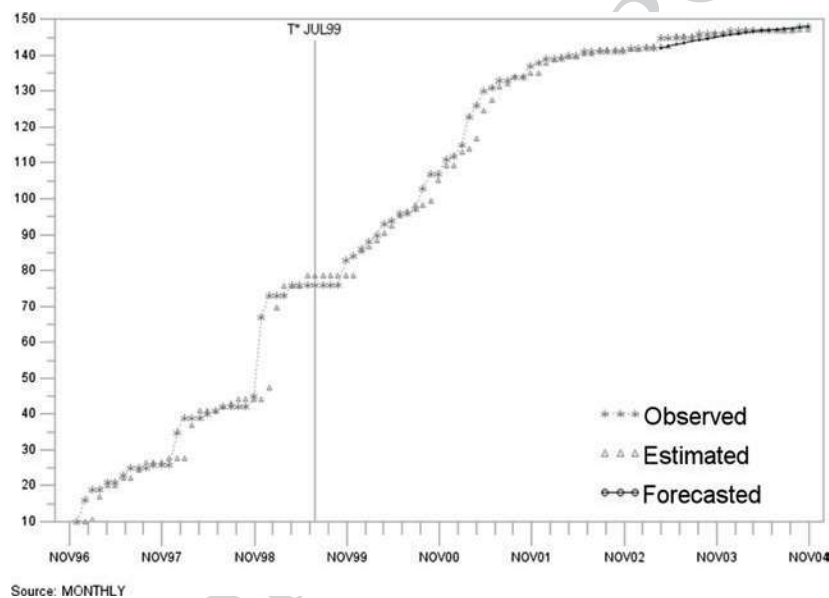


Fig. 6.4 Swiss Destination Managements Organizations (DMOs) (*Source* Authors' computation)

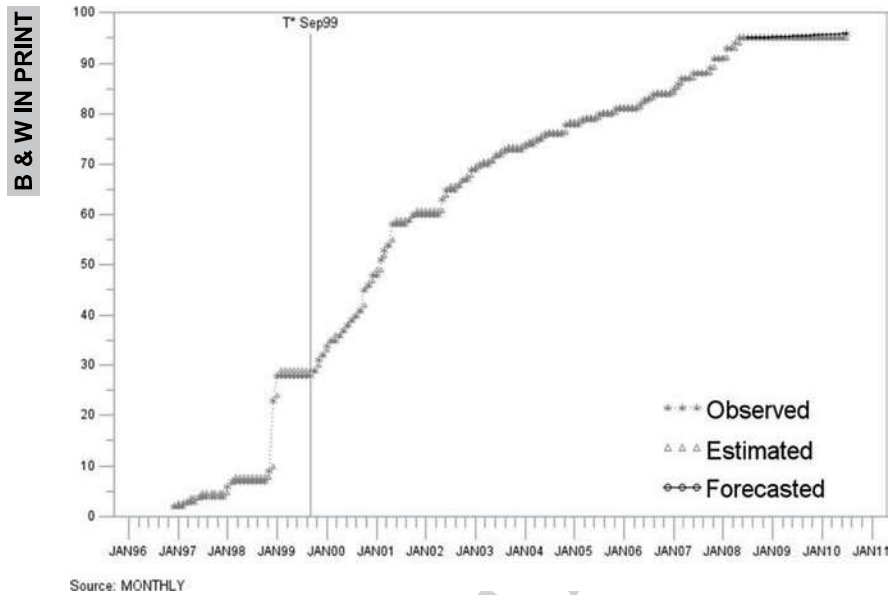


Fig. 6.5 Austrian DMOs (*Source* Authors' computation)

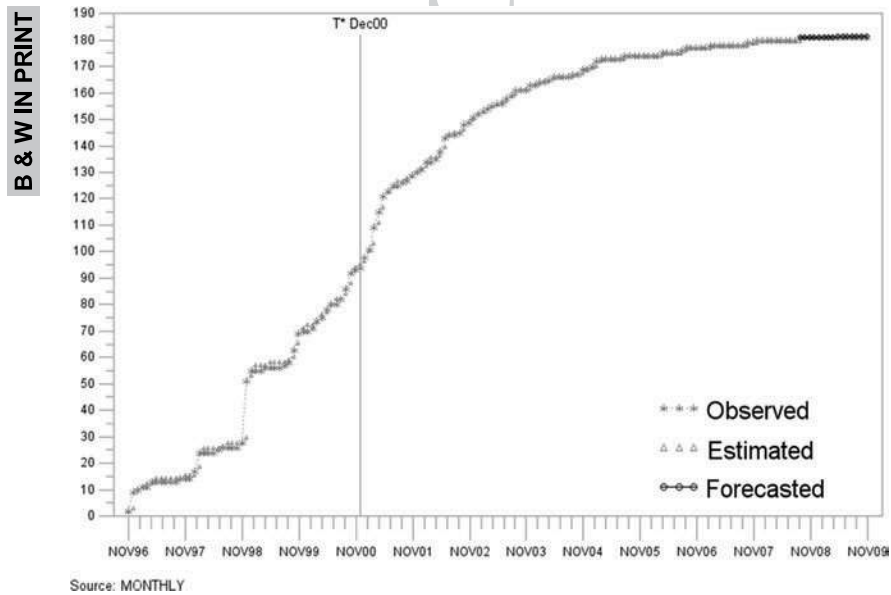


Fig. 6.6 German DMOs (*Source* Authors' computation)

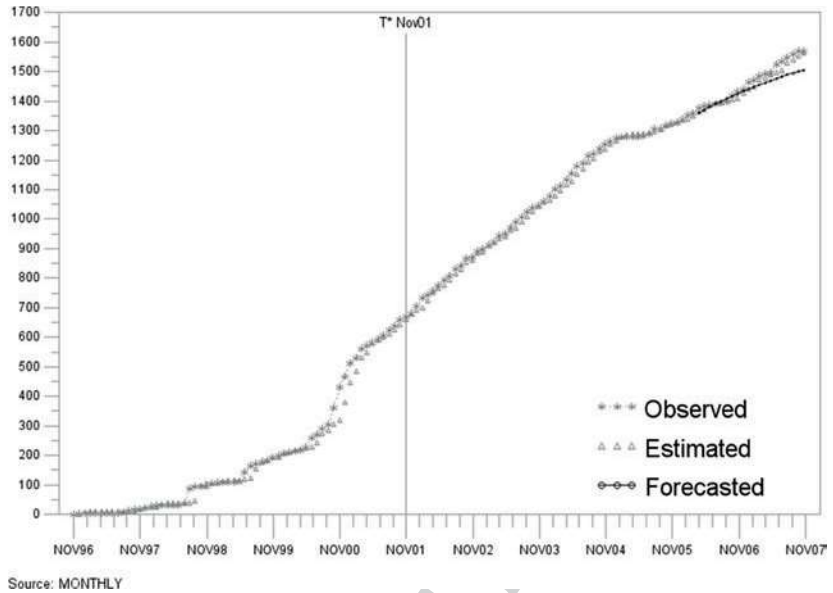


Fig. 6.7 Swiss restaurants (*Source* Authors' computation)

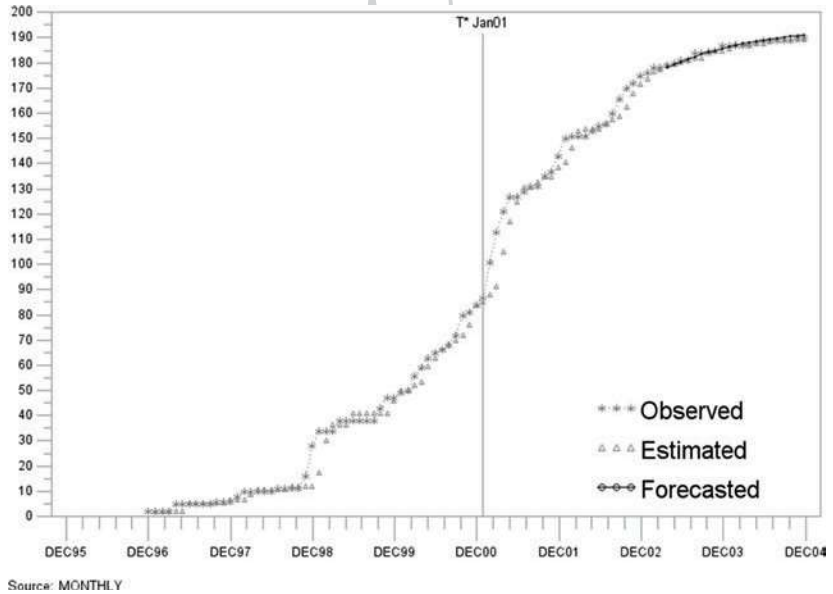


Fig. 6.8 Swiss cable cars companies (*Source* Authors' computation)

B & W IN PRINT

B & W IN PRINT

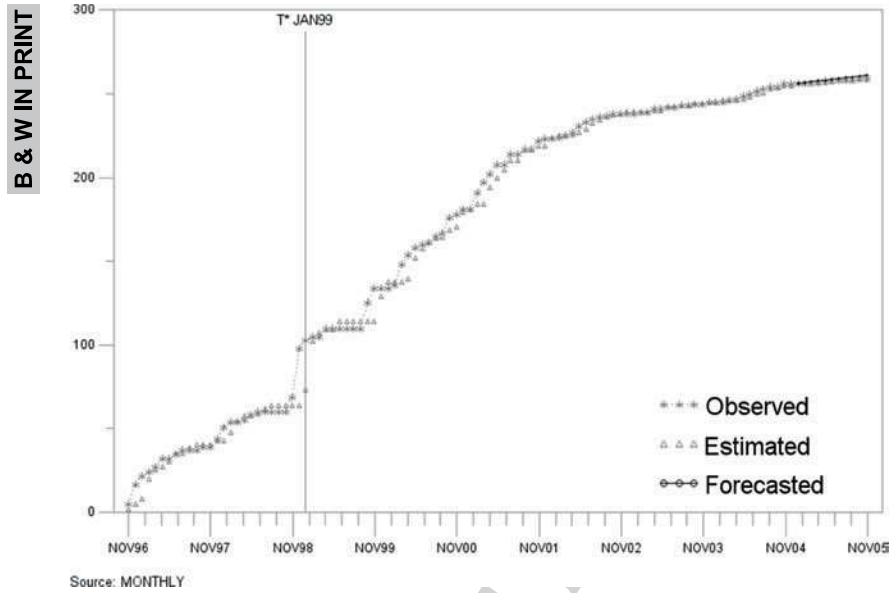


Fig. 6.9 International hotel chains (*Source* Authors' computation)

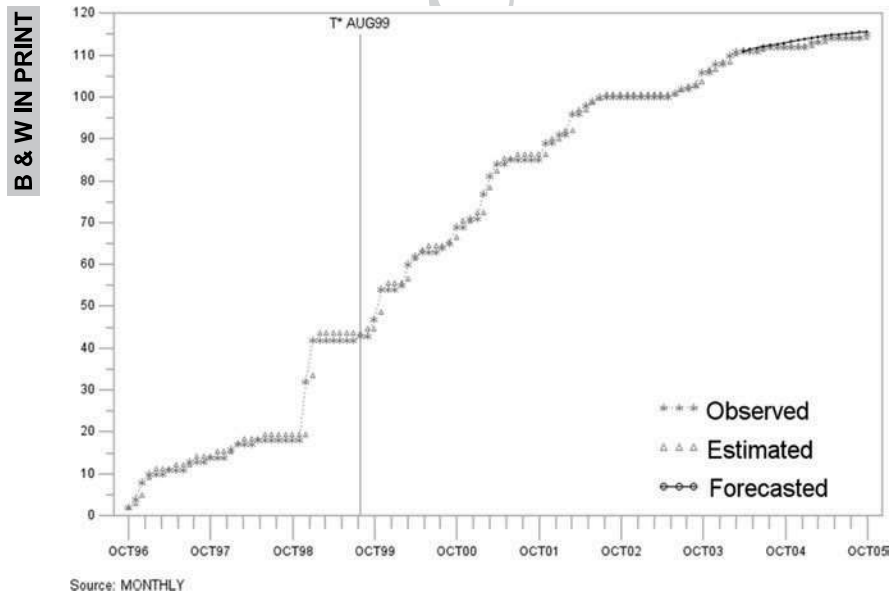


Fig. 6.10 European tour operators (TOs) (*Source* Authors' computation)

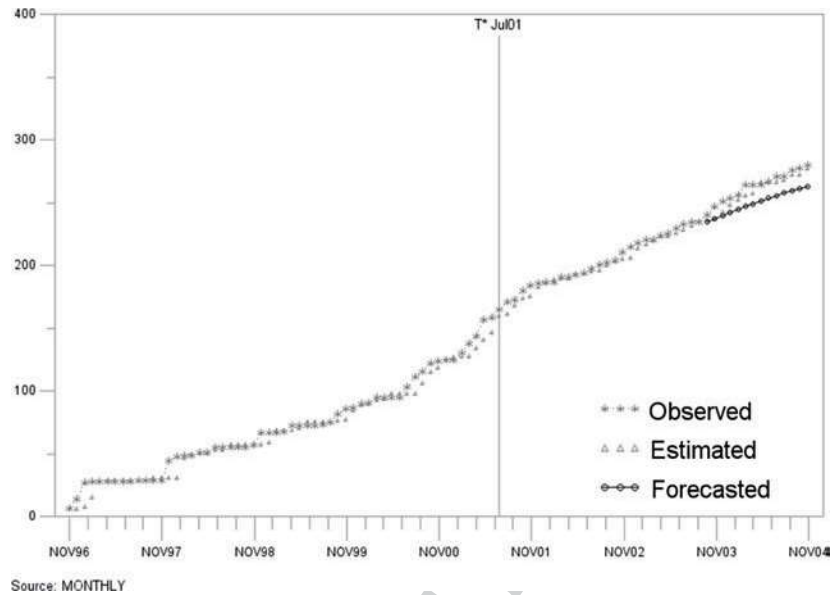


Fig. 6.11 Malaysian hotels (*Source* Authors' computation)

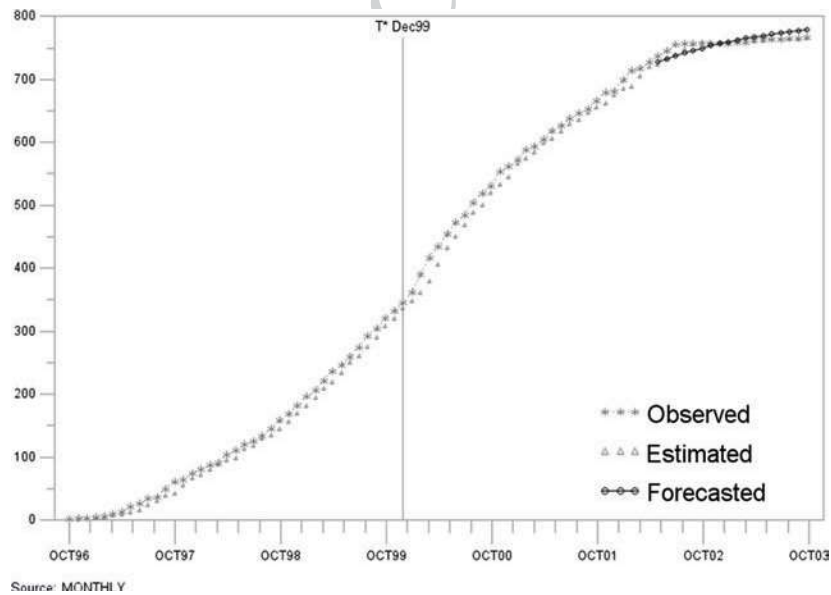


Fig. 6.12 Swiss non-affiliated hotels (*Source* Authors' computation)

B & W IN PRINT

B & W IN PRINT

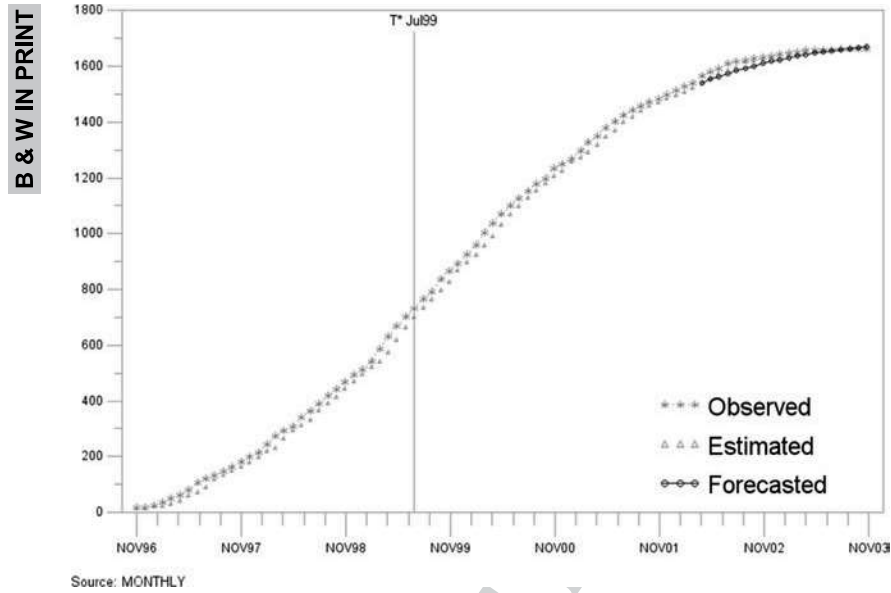


Fig. 6.13 Swiss affiliated hotels (*Source* Authors' computation)

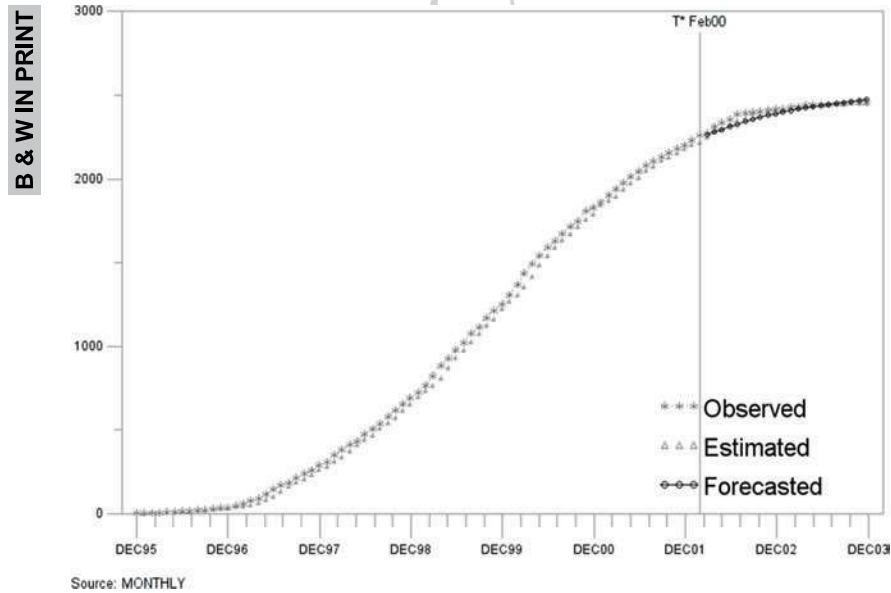


Fig. 6.14 Swiss hotels (*Source* Authors' computation)

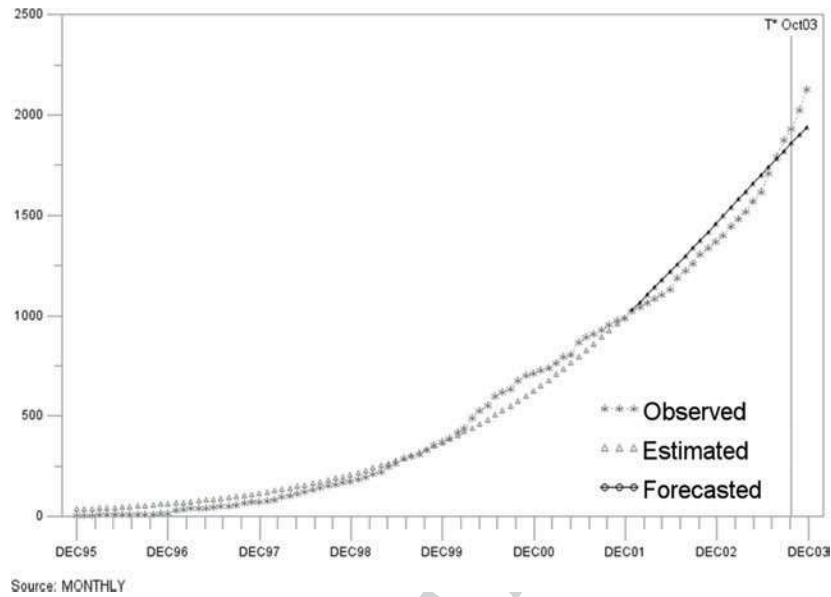


Fig. 6.15 Swiss *gasthouses* (*Source* Authors' computation)

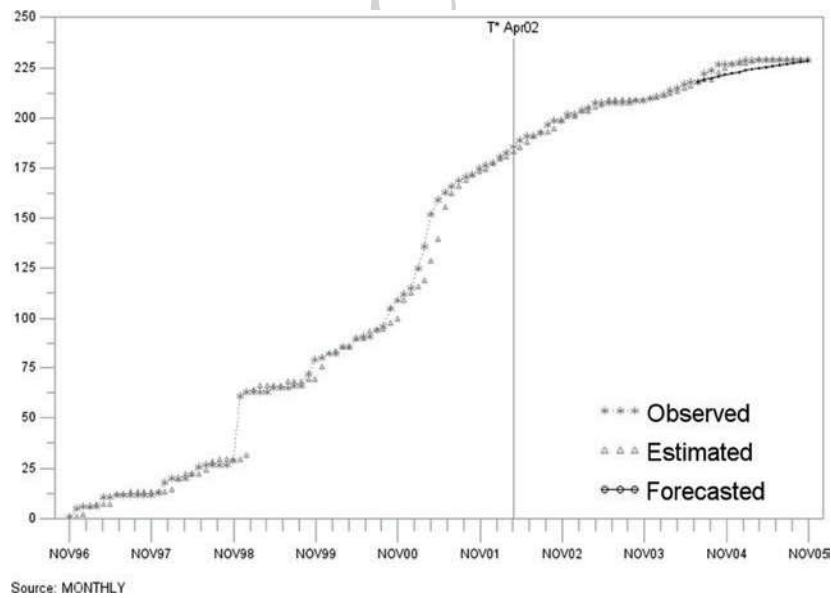


Fig. 6.16 Swiss travel agencies (*Source* Authors' computation)

B & W IN PRINT

B & W IN PRINT


**Table 6.3** Residual and forecasting error statistics

	Residual error statistics: estimation model Bass reported in Table 6.2				Forecasting errors statistics-24 last months				
	MSE	Root MSE	Adj R ²		Range of the sample (24 last observations)	Mean Abs Error	Mean Abs % Error	RMS Error	RMS % Error
DMO CH	7.3937	2.7191	0.9966		MAR2003–FEB2005	0.6861	0.4700	0.9199	0.6322
DMO AU	1.754	1.3244	0.9981		AUG2008–JUL2010	0.4610	0.4852	0.5309	0.5589
DMO GE	4.6745	2.162	0.9987		SEP2008–AUG2010	0.3447	0.1904	0.3856	0.213
Restaurant CH	105.5	10.2729	0.9997		MAR2006–FEB2008	26.4983	1.7302	34.5525	2.228
Cable car CH	6.022	2.454	0.9989		MAR2003–FEB2005	0.8279	0.4426	0.9744	0.5199
Hotel chain	12.3322	3.5117	0.9983		DEC2004–NOV2006	1.1229	0.4312	1.305	0.5013
European tour operators	4.0498	2.0124	0.9973		MAR2004–FEB2006	0.9542	0.8447	1.1094	0.9831
Malaysian hotel	7.5925	2.7554	0.9993		SEP2003–AUG2005	10.1686	3.7449	10.8393	3.9852
No affiliated hotel CH	14.2494	3.7748	0.9998		FEB2002–MAR2004	7.2274	0.9499	8.4162	1.1046
Hotelleriesuisse CH	29.2468	5.408	0.9999		MAR2002–FEB2004	11.5324	0.7077	13.5408	0.8346
Hotel CH	54.6344	7.3915	0.9999		FEB2002–JAN2004	16.1906	0.6734	19.8001	0.8263
Gasthouses CH	3922	62.6261	0.9892		JAN2002–DEC2003	67.9966	4.5786	78.837	5.0285
Travel agencies CH	13.1253	3.6229	0.9981		JUN2004–MAY2006	2.2008	0.9664	2.7593	1.2122

Source: Authors' calculations

REFERENCES

- 312
- 313 Andrés, L., D. Cuberes, M. Diouf, and T. Serebrisky. 2010. The Diffusion of
- 314 the Internet: A Cross-Country Analysis. *Telecommunications Policy* 34 (5–6):
- 315 323–340.
- 316 Bass, F.M. 1969. A New Product Growth Model for Consumer Durables.
- 317 *Management Science* 15 (5): 215–227.
- 318 Bemmaor, A.C. 1994. Modeling the Diffusion of New Durable Goods:
- 319 Word-of-Mouth Effect Versus Consumer Heterogeneity. In *Research*
- 320 *Traditions in Marketing*, ed. G. Laurent, G.L. Lilien, and B. Pras, 201–223.
- 321 Boston, MA: Kluwer Academic Publish Group.
- 322 Bemmaor, A.C., and J. Lee. 2002. The Impact of Heterogeneity and
- 323 Ill-Conditioning on Diffusion Model Parameter Estimates. *Marketing Science*
- 324 21 (2): 209–220.
- 325 Chatterjee, R., and J. Eliashberg. 1990. The Innovation Diffusion Process in
- 326 a Heterogeneous Population: A Micromodeling Approach. *Management*
- 327 *Science* 36 (9): 1057–1079.
- 328 Easingwood, C.J., V. Mahajan, and E. Muller. 1983. *A Nonuniform Influence*
- 329 *Innovation Diffusion Model of New Product Marketing Science* 2 (3): 273–295.
- 330 Hashim, N.H., M. Scaglione, and J. Murphy. 2012. Modelling and Comparing
- 331 Malaysian Hotel Website Diffusion. In *Information and Communication*
- 332 *Technologies in Tourism 2012, Proceedings of ENTER 2012*, ed. M. Fuchs,
- 333 F. Ricci, and L. Cantoni, 167–178. Wein and New York: Springer-Verlag.
- 334 Jensen, R. 1982. Adoption and Diffusion of an Innovation of Uncertain
- 335 Profitability. *Journal of Economic Theory* 27 (1): 182–193.
- 336 Jeyaraj, A., J.W. Rottman, and M.C. Lacity. 2006. A Review of Predictors, AQ9
- 337 Linkages, and Biases in IT Innovation Adoption. *Journal of Information*
- 338 *Technology* 21: 1–23.
- 339 Kim, S. 2011. The Diffusion of the Internet: Trend and Causes. *Social Science*
- 340 *Research* 40: 602–613.
- 341 Klimck, K., M. Scaglione, R. Schegg, and R. Matos-Wasem. 2012. Marketing
- 342 and Sustainable Tourism in Alpine Destinations. In *New Challenges for*
- 343 *Tourism Promotion: Tackling High Competition and Multimedia Changes*, vol.
- 344 6, ed. P. Keller and C. Laesser, 155–169. Berlin: Erich Schmidt Verlag.
- 345 Mahajan, V., E. Muller, and R. Srivastava. 1990. Determination of Adopter
- 346 Categories by Using Innovation Diffusion Models. *Journal of Marketing*
- 347 *Research* XXVII (February): 37–50.
- 348 Mahler, A., and E.M. Rogers. 1999. The Diffusion of Interactive
- 349 Communication Innovations and the Critical Mass: The Adoption of
- 350 Telecommunications Services by German Banks. *Telecommunications Policy*
- 351 23: 717–740.
- 352 Meade, N., and T. Islam. 2006. Modelling and Forecasting the Diffusion of
- 353 Innovation-A 25-Year Review. *International Journal of Forecasting* 22 (3): 519–545.

Layout: A5 HuSSci	Book ID: 490655_1_En	Book ISBN: 978-3-030-40601-1	
Chapter No.: 6	Date: 27 March 2020 18:35	Page: 23/23	

- Murphy, J., N.H. Hashim, and P. O'Connor. 2007. Take Me Back: Validating the Wayback Machine. *Journal of Computer-Mediated Communication* 13 (1). <http://jcmc.indiana.edu/vol13/issue11/murphy.html>.
- O'Connor, P. 1999. *Electronic Information Distribution in Tourism and Hospitality*. Wallingford: CAB international.
- O'Connor, P., and J. Murphy. 2004. A Review of Research on Information Technology in the Hospitality Industry. *International Journal of Hospitality Management* 23 (5): 473–484.
- Oren, S.S., and R.G. Schwartz. 1988. Diffusion of New Products in Risk-Sensitive Markets. *Journal of Forecasting* 7 (4): 273–287.
- Rogers, E.M. 1962. *Diffusion of Innovations*, 1st ed. New York: The Free Press.
- Rogers, E.M. 2003. *Diffusion of Innovations*, 5th ed. New York: The Free Press.
- Sabre. n.d. *The Sabre Story*. Retrieved from <https://www.sabre.com/website:https://www.sabre.com/files/Sabre-History.pdf>.
- SAS Institute Inc. 2011. *SAS/STAT® 9.22 User's Guide*. Cary, NC: SAS Institute Inc.
- Scaglione, M. Forthcoming. The Diffusion of Information and Communication Technologies (ICT) in the Tourism Sector. In *Handbook of E-Tourism*, ed. Z. Xiang, M. Fuchs, U. Gretzel, and W. Höpken. Cham: Springer International Publishing.
- Scaglione, M., C. Johnson, and J.-P. Trabichet. 2010. How the Swiss Tourism Sector Is Managing the Change to Web 2.0. In *Managing Change in Tourism Creating Opportunities—Overcoming Obstacles*, ed. P. Keller and T. Bieger, 101–118. Berlin: ERICH SCHMIDT VERLAG.
- Scaglione, M., and R. Schegg. 2015. The Case of Switzerland During the Last 20 Years. In *Tourism and Leisure: Current Issues and Perspectives of Development*, ed. H. Pechlaner and E. Smeral, 175–202. Berlin, Germany: Springer Verlag.
- Scaglione, M., R. Schegg, T. Steiner, and J. Murphy. 2004. The Diffusion of Domain Names by Small and Medium-Sized Swiss Hotels. In *The Future of Small and Medium Sized Enterprises in Tourism 54th AIEST Congress*, vol. 46, ed. P. Keller and T. Bieger, 259–271. St. Gallen: International Association of Scientific Experts. in Tourism.
- Steiner, T. 1999. Distributed Software Agents for WWW-Based Destination Information System. PhD Thesis. Lausanne: University of Lausanne.
- Van den Bulte, C., and S. Stremersch. 2004. Social Contagion and Income Heterogeneity in New Product Diffusion: A Meta-Analytic Test. *Marketing Science* 23 (4): 530–544.
- Zhu, K., and K.L. Kraemer. 2005. Post Adoption Variations in Usage and Value of E-Business by Organizations: Cross-Country Evidence from the Retail Industry. *Information Systems Research* 16 (1): 61–84.

Author Query Form

Book ID: **490655_1_En**
Chapter No: **6**

Please ensure you fill out your response to the queries raised below and return this form along with your corrections.

Dear Author,
During the process of typesetting your chapter, the following queries have arisen. Please check your typeset proof carefully against the queries listed below and mark the necessary changes either directly on the proof/online grid or in the ‘Author’s response’ area provided

Query Refs.	Details Required	Author’s Response
AQ1	Please check and confirm if the authors and their respective affiliations have been correctly identified. Amend if necessary.	
AQ2	The opening parenthesis does not have a corresponding closing parenthesis in the sentence ‘Since that moment, information ...’. Please insert the parenthesis in the appropriate position.	
AQ3	Please check and confirm the edits made in the citations ‘Bemmar (1994, p. 203)’ in the sentence ‘Rather than using the timing’...’.	
AQ4	The closing quote does not have a corresponding opening quote in the sentence ‘Analysis of the q/p ratio gives ...’. Please insert the quote in the appropriate position.	
AQ5	Kindly note that the term ‘p. 534’ has repeated in citation so we deleted repeated occurrence in the sentence ‘At the individual level ...’. Please check and confirm.	
AQ6	Please provide a definition for the significance of asterisk ‘*, **’ in the ‘Table 6.1’.	

Query Refs.	Details Required	Author's Response
AQ7	The closing parenthesis does not have a corresponding opening parenthesis in the column head "Observedpenetration ..." in Table 6.1. Please insert the parenthesis in the appropriate position.	
AQ8	The closing parenthesis does not have a corresponding opening parenthesis in the column head "Final market size respect to 1 ..." in Table 6.2. Please insert the parenthesis in the appropriate position.	
AQ9	Please provide a definition for the significance of 'letter a' in the "Table 6.2".	
AQ10	Reference 'Jeyaraj et al. (2006)' is given in the list but not cited in the text. Please cite this in text or delete from the list.	