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Seiritsu Ogura and Takehiko Hagino

### 9.1 Introduction

In Japan, drugs account for an enormous proportion of health care costs. Of the five major developed countries listed in table 9.1, Japan and France have the highest absolute per-patient cost of drugs, spending three times as much as England, twice as much as the United States, and one and a half times as much as Germany. Japan consistently ranks at the top in terms of the share of drugs in total health care costs, spending more than 20 percent on drugs according to the Organization for Economic Cooperation and Development (OECD) Health Data (1998, table 9.2).

According to the 1996 Survey on Socialized Medicine (Ministry of Health and Welfare 1994–1996)<sup>1</sup> injections account for as great a share of inpatient costs as do surgeries (9.5 percent and 9.4 percent, respectively), while drugs and injections together consume 39.6 percent of outpatient health care costs. Although drugs are a very important component of to-tal health care expenditures in Japan, their importance seems to be heavily concentrated in the outpatient care of elderly patients. In inpatient care, drugs account for only 12.8 percent of the difference in average costs between the elderly and the rest of the population. In outpatient care, drugs account for 56 percent of this difference. Given the rapid aging of the Japanese population, it is clear that if we do not find some way to control

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<sup>1.</sup> This survey, based on approximately 300,000 reimbursement claims submitted by health care providers, is published annually by the Ministry of Health and Welfare.

	National Health Care Expenditures (yen per capita)	Outpatient Drug Expenditures (yen per patient)	Inpatient Drug Expenditures (yen per patient)	Share of All Drugs in National Health Care Expenditure (%)
France	253,680	43,375	50,375	19.9
Germany	224,420	32,195	38,283	17.1
England	110,625	16,341	18,153	16.4
United States	312,755	23,076	35,418	11.3
Japan	195,217	43,533	57,589	29.5

#### Table 9.1 International Health Care and Drug Expenditures, 1993

Source: Illustrated White Paper on Health Insurance (1998).

*Note:* Data for England are from 1992. Exchange rates: 23 yen/franc, 79 yen/deutsche mark, 231 yen/pound, and 122 yen/dollar.

Table 9.2	Shar	e of Drug	s in Natior	al Health	Care Cost	s, 1990–19	998 (%)	
	1990	1991	1992	1993	1994	1995	1996	1997
France	16.7	16.7	16.6	16.8	16.6	16.7	17.0	16.7
Germany	14.2	14.3	14.2	12.4	12.3	12.3	12.7	12.6
England	13.8	14.0	14.5	15.3	15.3	15.9	16.5	17.3
United States	8.6	8.5	8.5	8.4	8.4	8.6	8.8	7.8
Japan	21.4	22.8	21.9	22.1	20.9	20.2	20.8	20.0

Source: 1998 OECD Health Data

the cost of drugs, we will not be able to control health care expenditures in the twenty-first century.

Why do we spend so much on drugs? The first place to look for an answer is in the enormous distortions generated by drug price regulations during the last five decades. Although the entire health care sector lies outside the realm of the market economy (the government sets comprehensive reimbursement prices for individual health care services in an attempt to control every possible aspect of health care service delivery), these regulations on drug prices are very peculiar. Regulated drug prices vary by brand name even if they are chemically identical. In general, when the government purchases any other good or service it must observe a set of very stringent procurement procedures designed to assure the lowest possible price. With drugs, physicians are allowed to use more expensive brandname drugs when cheaper alternatives are available, even though they are supposed to be acting as agents of the government.

Second, the government sets reimbursement (retail) prices on the basis of market (wholesale) prices from almost eighteen months earlier, and these retail prices are then fixed for the following two years. If the market functioned normally, such regulation would be a source of enormous risk for health care providers. In other words, the market structure must be very peculiar for such a system to have functioned without driving a large number of providers to bankruptcy.

Third, drug markets are very tightly regulated and protected by extremely high barriers to new entry. The regulations concerning the introduction of new drugs work as prohibitive barriers to entry, as they are both very time consuming and costly. At the same time, these high barriers make it easy for insiders to set noncompetitive prices and to build cozy relationships with regulators.

In this paper, we present evidence regarding the effects of price distortions on resource allocation in the health care sector. Our estimates indicate that the magnitude of these effects exceeds 20 percent, and may be as high as 50 percent, of drug costs. We then show that the government's attempts to control drug prices directly are at best ineffective, as they have been offset by drug-switching effects in most drug groups. These drugswitching effects are in turn induced by the built-in profit margins for "new" drugs, which are generously priced by regulators. We base our conclusions on statistical analyses of the first comprehensive microdata set compiled in Japan.

The rest of the paper proceeds as follows: In section 9.2, we review the drug price controls of the last decade and classify the economic inefficiencies associated with these controls; in section 9.3, we analyze the drug selection behavior of physicians; in section 9.4, we present a decomposition analysis of drug costs; in section 9.5, we discuss various reform proposals currently under consideration; section 9.6 concludes.

### 9.2 Drug Pricing in the Japanese Health Care System

As all drugs are privately produced, health care service providers must purchase all necessary drugs through the market. Trading is free for almost all drugs, and market prices are formed on the basis of reimbursement prices. In effect, the reimbursement prices are the drugs' retail prices and market prices are the drugs' wholesale prices, generating profit margins for each drug.

### 9.2.1 Formula for Drug Price Revisions

The Ministry of Health and Welfare (MHW) updates the list of reimbursement prices for individual drugs approximately every two years using its *Survey of Drug Prices*.<sup>2</sup> The current list of drugs contains almost 14,000

<sup>2.</sup> In the *Survey of Drug Prices*, the government collects data on the actual purchase prices of individual drugs from health care providers. The survey's objectivity, however, is somewhat weakened because it is announced well in advance and data are collected for just one month (usually September).

different drugs. For nearly ten years, the government has been using the following formula to revise the reimbursement prices:<sup>3</sup>

(1) 
$$Y_t = X_{t-1} + r_t Y_{t-1},$$

where

 $Y_t$ : New drug price,

 $X_{t-1}$ : Average market price in the Survey of Drug Prices,

 $r_t$ : Reasonable zone factor ( $0 < r_t < 1$ ), and

 $Y_{t-1}$ : Old drug price.

Once a new price is set, it is fixed for the next two years. Hence, at least in theory, providers take on the considerable risk of market price variation in drug prices. In the last ten years of practice, however, once the new prices are announced almost all market prices have continuously declined. As a result, providers enjoy positive profit margins on almost all drugs and so have an economic incentive to sell as many drugs as possible to their patients. There are two reasons for this: the reasonable zone factor and the generous price setting for "new" drugs.

# 9.2.2 Drug Pricing Rules

If we subtract old drug prices from both sides of equation (1), we obtain

(2) 
$$Y_t - Y_{t-1} = -(Y_{t-1} - X_{t-1}) + r_t Y_{t-1}$$

By dividing both sides by old drug prices, we obtain the revision formula in proportion terms

(3) 
$$y_{t-1} = -\pi_{t-1} + r_t$$

where  $y_{t-1}$  is the rate of change of the reimbursement price of the drug and  $\pi_{t-1}$  is the discount rate the firm offered to providers in the previous period.

(4) 
$$y_{t-1} = \frac{(Y_t - Y_{t-1})}{Y_{t-1}}$$

(5) 
$$\pi_{t-1} = \frac{(Y_{t-1} - X_{t-1})}{Y_{t-1}}$$

In the absence of the reasonable zone factor, equation (5) implies that if a firm offers a discount to promote the sale of a drug, that drug's price is permanently lowered. Such a prospect should make the firm hesitant to offer a discount. With the reasonable zone factor, however, the firm can offer price discounts without endangering future profitability, provided that the discount rate remains within the range of the reasonable zone factor.

<sup>3.</sup> Strictly speaking, this formula has only been adopted for all drugs since 1992. Prior to that time, "bulk-line" formulas were used.

Recently, however, the government has been forced to rapidly reduce both drug prices<sup>4</sup> and reasonable zone factors in an attempt to control health care costs, particularly those of the elderly. The zone factor dropped from 15 percent in 1992 to 13 percent in 1994, 11 percent in 1996, 10 percent in 1997, and 5 percent in 1998. Starting in 1997, an even lower rate was applied to "new" drugs with chemical components similar to existing drugs: 8 percent in 1997 and 2 percent in 1998.

# 9.2.3 Assessing Price Regulations: Decomposition of the Variation in the Cost of Drugs

As we have shown, Japanese drug price regulation has a built-in price reduction mechanism. If there were no change in quantity used, price would change at exactly the same rate as cost. The overall cost of drugs in public health insurance, however, follows two-year cycles, showing decreases in the years of price revisions and increases in the years following price revisions (table 9.3). Many (Ikegami et al. 1998; Ogura 1996, 1998) believe that this pattern is explained by the combination of two trends working against the cut in drug prices: increasing reliance on multiple drugs and overall switching to more expensive drugs. Some add a third trend of switching to *new* expensive drugs. As far as we know, however, no one has quantified these trends using comprehensive drug-usage data.

### Five-Factor Decomposition

We decompose the variation in drug costs using the following five factors: (a) prescription probability factor; (b) drug-switching factor; (c) price regulation factor; (d) daily quantity factor; and (e) number of days prescribed factor. All drugs are classified at the three-digit level. The perpatient average cost of drugs in group k in period t is denoted  $v_k(t)$ ;  $v_k(t)$  is the product of  $\Gamma_k(t)$ , the probability that any drug in group k is prescribed; and  $\Omega_k(t)$  is the per-patient cost of drugs in group k (given that at least one drug in the group has been prescribed). Therefore, we have

$$V(t) = \sum_{k=1}^{n} v_k(t)$$
$$v_k(t) = \Gamma_k(t)\Omega_k(t)$$

The per-patient cost,  $\Omega_k(t)$ , is in turn the product of four factors (summed over all *i*):  $s_i(t)$ , the probability of selecting drug *i* in group *k*;  $p_i(t)$ , the reimbursement price of drug *i*;  $x_i(t)$ , the daily quantity of drug *i*; and  $d_i(t)$ , the number of days that drug *i* is used. Or,

$$\Omega_k(t) = \sum_i s_i(t) p_i(t) x_i(t) d_i(t).$$

4. For the officially announced changes in drug prices, see table 9.3.

Table 9.3	Changes in Drug Prices, 1990–1996	Prices, 1990-	1996						
	Chance in		Drug Spending, All Patients (yen)	All Patients (	yen)	Ū	Drug Spending, Outpatients Only (yen)	tpatients Onl	y (yen)
Year	Drug Prices (%)	Drugs	Injections	Total	Change (%)	Drugs	Injections	Total	Change (%)
1990	-9.2	5,359	2,027	7,386	-5.7	4,945	634	5,579	-1.8
1991		5,839	2,100	7,939	7.5	5,405	688	6,093	9.2
1992	-8.1	5,760	2,074	7,834	-1.3	5,380	688	6,068	-0.4
1993		5,863	2,084	7,947	1.4	5,493	805	6,298	3.8
1994	-6.6	5,564	1,789	7,353	-7.5	5,255	689	5,944	-5.6
1995		5,841	1,903	7,745	5.3	5,546	758	6,304	6.1
1996	-6.8	5,433	1,823	7,256	-6.3	5,156	760	5,916	-6.2
1989 - 1996	-30.7				4.1				7.9
Source: Survey	Source: Survey of National Medical C	are Insurance	dedical Care Insurance Services 1990–1996 (MHW 1994–1996)	96 (MHW 19	94–1996)				

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Note that the sum of the  $s_i(t)$  terms, each of which represents the probability of selecting drug *i* conditional on prescribing some drug in group *k*, equals one.

Given the above formulation, the variation in  $v_k(t)$  is determined by the variation in the five component variables. The proportional rate of change in  $v_k(t)$  is given by

$$\begin{split} \frac{\Delta v_k(t)}{v_k(t)} &= \frac{\Delta \Gamma_k(t)}{\Gamma_k(t)} + \frac{\Delta \Omega_k(t)}{\Omega_k(t)} = \frac{\Delta \Gamma_k(t)}{\Gamma_k(t)} + \frac{\sum_i \Delta s_i(t) p_i(t) x_i(t) d_i(t)}{\sum_i s_i(t) p_i(t) x_i(t) d_i(t)} \\ &+ \frac{\sum_i s_i(t) \Delta p_i(t) x_i(t) d_i(t)}{\sum_i s_i(t) p_i(t) x_i(t) d_i(t)} + \frac{\sum_i s_i(t) p_i(t) \Delta x_i(t) d_i(t)}{\sum_i s_i(t) p_i(t) x_i(t) d_i(t)} \\ &+ \frac{\sum_i s_i(t) p_i(t) x_i(t) \Delta d_i(t)}{\sum_i s_i(t) p_i(t) x_i(t) d_i(t)}. \end{split}$$

In words, we offer the following explanation:

1. The first term is the rate of change in the probability of prescribing drugs in group k. The trend toward relying on more groups of drugs is found by summing the first term over all the drug groups.

2. The second term is the drug-switching effect within group k, or the rate of change in the per-patient cost of drugs as a result of changes in drug selection. The trend of switching to expensive drugs is observed directly in the second term.

3. The third term is the price regulation effect in group k, or the rate of change in the per-patient cost of drugs as a result of changes in regulated drug prices. This is the government's control variable, and the sum of the third term over all the drug groups should be close to the rate the government announces.

4. The fourth term is the rate of change in the per-patient cost of drugs as a result of changes in daily quantities.

5. The fifth term is the rate of change in the per-patient cost of drugs as a result of changes in the number of days for which drugs are prescribed.

### Data: Modification and Errors

We use the 1994, 1995, and 1996 *Survey of Drugs* for our analysis (MHW). The data are described in detail in section 9.3.2. There are two groups of claims that are dropped from the original data sets. First, we exclude all claims submitted by drug stores because they do not include patient identification numbers. Second, we exclude observations above the 99th percentile in each three-digit group in terms of daily quantities. We make this exclusion because of the extremely large variance in daily quantities for some groups. The descriptive statistics for our sample are given in table 9.4.

		Sample Size		Weig	Weighted Sample Size	Size	Num	Number of Drugs	Drugs	0	Cost Share	9	Averag Da	Average Number of Days Taken	ber of n	Averag	Average Cost per Day (yen)	r Day
Group	1994	1995	1996	1994	1995	1996	1994	1995	1996	1994	1995	1996	1994	1995	1996	1994	1995	1996
114	31,577	25,259	23,105	3,655,202	3,407,630	3,186,121	177	169	174	1.97	1.88	1.75	7.8	8.0	7.6	147	158	157
117	8,977	7,889	7,399	624,664	613,802	626,488	168	160	181	1.16	1.01	1.12	37.3	37.7	40.9	107	100	95
119	10,547	8,937	7,450	804,231	815,196	723,704	16	16	20	3.06	3.04	2.68	26.9	28.6	28.7	304	301	278
131	20,384	17,359	17,521	2,943,641	2,913,736	2,936,797	141	138	154	2.22	2.23	2.59	2.4	2.5	2.5	688	717	765
212	4,893	4,045	3,700	616,395	574,248	595,687	71	74	83	1.65	1.48	1.67	27.1	27.3	27.9	213	218	216
214	9,478	8,486	8,165	1,177,132	1,283,977	1,303,849	121	120	145	2.57	2.77	3.03	31.4	33.3	33.2	150	149	151
217	16,474	14,555	13,808	1,722,539	1,779,782	1,740,082	93	95	105	3.04	3.03	3.06	27.3	27.7	26.4	139	142	143
218	10,435	9,622	7,860	1,522,350	1,676,518	1,396,209	32	32	34	3.88	4.15	3.70	26.6	27.7	27.6	205	206	207
219	16,483	13,413	10,495	1,407,114	1,305,332	1,013,397	95	95	109	4.00	3.80	3.33	27.8	28.8	28.2	219	233	251
232	29,016	24,270	20,800	3,143,325	3,117,989	2,661,715	111	108	132	4.14	3.95	3.20	26.1	25.7	25.0	108	113	104
239	10,709	9,066	8,284	1,092,362	1,067,588	1,022,061	41	42	52	0.97	1.04	0.90	13.6	12.4	12.5	139	183	152
241	757	396	463	61,739	43,055	50,419	19	19	18	1.05	1.24	1.27	2.7	2.1	2.0	13,582	32,402	26,989
249	5,329	4,410	3,852	589,651	568,551	485,626	51	4	49	1.93	1.98	2.27	12.5	11.7	9.7	561	686	1,036
259	4,664	4,662	4,886	380,981	465,509	491,760	20	20	26	1.23	1.54	1.66	24.4	27.0	27.4	285	282	265
264	64,118	55,264	54,666	7,828,626	7,903,908	7,735,899	155	149	169	4.85	4.59	4.56	2.5	2.4	2.4	536	554	527
325	4,777	3,930	4,037	249,848	259, 122	267,570	76	72	80	1.82	1.67	2.19	14.0	13.9	15.8	1,116	1,072	1,122
331	30,507	26,147	27,181	1,993,999	2,189,348	2,317,219	6L	78	84	1.73	1.90	2.01	7.0	7.5	7.3	265	266	257
339	6,589	5,380	6,171	681,982	645,230	797,465	22	21	27	4.13	3.42	4.48	27.5	27.3	28.4	473	447	426
399	15,234	13,124	12,696	1,249,896	1,287,442	1,352,312	69	71	83	6.89	7.48	7.58	10.1	10.0	9.6	1,172	1,331	1,258
422	1,844	1,541	1,369	229,891	236, 293	195,668	56	51	49	3.80	3.71	2.90	22.6	23.5	22.4	1,568	1,539	1,423
429	809	796	842	84,001	96,666	92,219	28	29	41	1.34	1.46	1.32	16.4	16.1	17.5	2,088	2,156	1,762
430	854	788	659	57,497	62,287	55,247	43	39	43	1.48	1.20	0.86	1.6	1.6	1.5	33,418	27,812	22,932
449	8,446	6,869	7,971	1,173,619	1,150,464	1,339,759	31	31	44	2.45	2.30	3.08	16.1	17.0	18.0	278	271	275
520	6,134	5,211	4,244	690,171	714,314	639,812	203	188	199	2.12	2.10	1.82	21.8	22.4	22.2	302	303	276
613	30,053	26,448	25,291	3,724,928	4,246,507	3,814,287	205	202	250	7.27	7.56	7.28	5.3	5.3	5.3	792	768	770
624	10,375	9,225	9,371	1,303,008	1,481,840	1,517,252	26	26	31	2.12	2.49	2.47	5.8	6.0	6.2	598	641	561
634	1,829	1,557	1,477	96,557	92,083	83,259	76	65	70	3.28	3.40	3.32	3.6	3.9	3.9	20,428	21,781	21,843
639	1,430	1,031	843	180,229	176, 196	127,060	28	24	29	0.84	1.46	1.50	2.3	2.4	2.9	4,310	8,032	8,735
721	6,998	6,077	4,864	618,810	632,830	517,710	89	84	101	3.58	3.46	2.47	1.7	1.7	1.7	7,119	7,364	6,165
5	6		-	-		1000												
Source:	Authors	Source: Authors computations based on surveys on drugs (MHW 1994–1996)	ased on su	rveys on drug	gs (MHW 19	94-1996).								•				

Drug Data Descriptive Statistics, by Three Digit Group

Table 9.4

Notes: The cumulative share of drugs in the listed drug groups is 80 percent. Data include drugs used in injections. One shot is a "day" for injections.

We still have to deal with two problems that plague any attempt to use indexes: what to do with drugs introduced into and removed from our sample. Drugs that are introduced for the first time in period t do not have data in period t-1, including price data. Drugs that are removed in period t do not have data in period t + 1, including price data. If we exclude both of them, we lose a significant part of total drug costs from our analysis, but if we include them we will no longer have an identity.

In the end, we have chosen inclusion with some modification. For the prescription factor and the drug-switching factor terms, we include both introduced and removed drugs. For the price regulation factor, daily quantity factor, and number of days prescribed factor terms, we include neither. This is the main reason for the nonnegligible residuals obtained in the decomposition analysis in certain cases.

### Results

The decomposition analyses have been carried out for all three-digit groups. Table 9.5 reports aggregated results (the sum of all drugs) with breakdowns for some provider and regional characteristics.

*Results for All Drugs.* Between 1994 and 1995, when there were no revisions of drug prices, the average cost of drugs increased by 5.1 percent. According to our decomposition, changes in the prescription probability increased the cost of drugs by 1.7 percent. The drug-switching effect increased the cost of drugs by 5.4 percent, while changes in daily quantities decreased the cost of drugs by 0.7 percent. Finally, changes in the number of days prescribed increased the cost of drugs by 1.5 percent. The errors therefore amount to -2.8 percent.

Between 1995 and 1996, when drug prices were revised, average drug costs fell by 9.7 percent. The reduction in drug prices decreased the cost of drugs by 7.7 percent, and changes in the prescription probability decreased the cost of drugs another 7.0 percent. The drug-switching effect increased the cost of drugs by 7.1 percent, changes in daily quantity increased the cost of drugs by 0.2 percent, and changes in the number of days prescribed decreased the cost of drugs by 0.1 percent.

*Price Controls and Drug-switching Effects.* The magnitudes of drugswitching effects are particularly important as an indicator of the effectiveness of drug price regulation. In table 9.6 we report the percentage changes in price in 1996 for ten major three-digit groups, as well as the drugswitching effects in 1995 and 1996. Our ten major groups account for almost 40 percent of the cost of drugs in 1995, and changes in their prices were enough to drive drug costs down by 3.3 percent. However, the drugswitching effects in these groups in 1995 and 1996 together drove drug costs

Table 9.5	Diug Cost	anation Decon	iipositions, 1.		inu 1996 19	<b>JU</b> (70)	
	Change in Total Drug Costs	Change in Prescription Probability	Drug Switching Effect	Change in Drug Prices	Change in Daily Quantity	Change in Number of Days Taken	Error
		1	994–1995				
Total	5.1	1.7	5.4		-0.7	1.5	-2.8
Inpatient care	5.1	0.8	5.4		-1.5	1.9	-1.6
Outpatient care	5.1	1.5	8.3		0.9	1.4	-7.0
Ownership							
National/public	4.9	4.7	4.4		2.7	3.1	-9.9
Private	4.2	2.5	6.8		5.1	1.5	-11.6
University	-11.3	9.9	22.3		-0.8	-1.5	-41.2
Clinic with beds	0.8	-1.6	4.4		-1.4	7.3	-8.0
Clinic without beds	6.8	4.3	6.3		9.1	3.6	-16.5
Number of beds							
0	4.2	3.8	6.1		5.9	2.3	-14.0
<200	5.4	3.4	3.0		-2.6	10.9	-9.3
>200	1.7	2.4	7.2		10.7	0.2	-18.9
Number of patients per physician							
<100	0.7	-0.9	2.0		15.4	2.1	-17.9
100-200	7.9	6.3	13.4		1.0	1.5	-14.4
>200 and <400	16.3	10.9	4.1		-0.2	6.5	-5.0
>400	9.0	12.4	51.7		-3.1	-0.9	51.1
Region							
Hokkaido	-0.2	14.1	9.0		-3.5	3.7	-23.6
Tohoku	8.1	8.9	13.0		-1.0	-1.4	-11.4
Kanto	-7.3	2.6	1.1		12.9	-3.3	-20.7
Hokuriku	-24.9	-6.4	27.5		-1.4	3.4	-48.0
Chubu-Tokai	8.2	4.4	28.4		-3.1	3.6	-25.3
Kinki	12.1	3.0	11.2		24.8	3.2	-30.1
Chugoku	-6.6	-0.8	8.3		-0.1	5.3	-19.2
Sikoku	10.2	18.3	13.6		-1.0	23.1	-43.8
Kyosyu	14.3	13.4	3.3		-3.5	7.6	-6.5
		1	995–1996				
Total	-9.7	-7.0	7.0	-7.7	0.2	-0.1	-2.2
Inpatient care	-9.8	-6.1	5.3	-7.2	-0.2	0.0	-1.6
Outpatient care	-9.6	-5.2	7.1	-7.9	1.0	-0.1	-4.5
Ownership							
National/public	-5.2	1.1	16.8	-7.2	-2.4	0.0	-13.6
Private	-16.8	-9.9	8.7	-8.1	0.7	0.4	-8.5
University	9.2	11.0	30.4	-5.4	-2.6	4.2	-28.4
Clinic with beds	-8.2	-3.7	10.6	-8.8	2.5	-1.2	-7.5
Clinic without beds Number of beds	-13.8	-7.2	6.3	-8.6	2.2	-1.5	-5.0
0	-14.5	-7.1	7.0	-8.6	1.1	-1.9	-5.0
<200	-5.9	-5.9	46.8	-8.7	2.3	-1.8	-38.6
>200	-3.7	2.8	6.7	-7.0	-4.6	4.4	-6.0

Drug Cost Variation Decompositions, 1994–1995 and 1995–1996 (%)

### Table 9.5

Table 9.5	(continued)						
	Change in Total Drug Costs	Change in Prescription Probability	Drug Switching Effect	Change in Drug Prices	Change in Daily Quantity	Change in Number of Days Taken	Error
Number of patients per physician							
<100	-7.3	-3.2	10.8	-7.4	-1.4	5.6	-11.7
100-200	-9.5	-1.2	8.8	-7.7	1.1	-1.0	-9.5
>200 and <400	-9.2	-4.2	6.8	-8.2	2.3	-0.3	-5.4
>400	-11.8	-4.5	4.2	-8.8	1.0	4.1	-7.7
Region							
Hokkaido	0.8	-0.2	15.6	-8.2	2.2	2.6	-11.2
Tohoku	-16.9	16.0	4.7	-7.9	-0.6	4.7	-33.8
Kanto	-2.4	27.2	18.2	-7.5	-0.8	4.8	-44.3
Hokuriku	5.3	461.3	9.5	-8.2	1.1	3.1	-461.5
Chubu-Tokai	-16.8	-7.0	2.7	-8.5	0.6	-0.8	-3.8
Kinki	-13.2	-2.3	9.1	-8.0	-1.1	4.0	-14.8
Chugoku	-5.3	3.4	5.4	-7.8	-1.0	0.7	-6.0
Sikoku	-1.9	13.1	44.4	-9.9	-2.9	2.0	-48.5
Kyusyu	-15.7	0.3	22.6	-7.7	-2.3	-0.7	-27.9

Source: Authors' computations based on surveys on drugs (MHW 1994-1996).

Table 9 5

(continued)

Table 9.6	Price Changes and Switching Effects for Selected Drug Groups (%)

	S	witching Effect			
Three-Digit Group	1995 Cost Share	1996 Price Change	1995	1996	Total
131	2.2	-3.9	4.52	12.01	16.53
214	2.8	-6.7	3.56	5.95	9.51
217	3.0	-7.9	5.55	9.28	14.83
218	4.2	-2.8	2.95	-0.91	2.04
219	3.4	-8.8	5.55	10.54	16.09
232	4.0	-9.4	5.22	2.52	7.74
264	4.6	-11.7	1.94	2.93	4.87
422	3.7	-5.6	-1.19	5.79	4.6
613	7.6	-11.4	-0.14	9.64	9.5
721	3.5	-11.7	14.96	-7.04	7.92
Total	38.83	-3.30	1.44	1.93	3.40

Source: Authors' computations based on surveys on drugs (MHW 1994-1996).

up by 3.4 percent. The relative magnitudes of the price changes and the switching effects vary widely across drug groups, but in the aggregate they are almost equal.

*Observations.* We offer the following conjectures based on these three years of data:

1. The drug-switching effects work to increase the cost of drugs at an annual rate of somewhere between 5 percent and 7 percent.

2. When there is no reduction in drug prices, the trend toward more prescriptions pushes up total drug costs by nearly 2 percent, but when drug prices are reduced, the trend is reversed and prescription probabilities contribute to a 7 percent decline in drug costs.

3. During the 1994–1996 period, the reduction in drug prices and prescription probability together barely offset the drug-switching effect.

# 9.3 Economic Incentives in Drug Choice

Many physicians strongly deny that their drug choices are affected in any way by economic incentives. Are their decisions purely based on medicine and truly free from economics? In this section we examine physician selection behavior using a formal model to analyze a comprehensive dataset on drugs chosen by physicians.

# 9.3.1 Selection Functions

Assume that a physician selects a drug out of *J* possible drugs using the following selection function:

$$u_{ik} = \beta'_k \mathbf{x}_i + \gamma'_k \mathbf{z}_k + \varepsilon_{ik},$$

where **x** is a vector of patient and provider characteristics, **z** is a vector of drug characteristics, and  $\varepsilon$  is the error term. The drug's profit margin is included as one of the elements of **z**, and the corresponding coefficient in  $\gamma$  gives the direction and size of the influence of the profit margin on the physician's selection. The physician selects drug *j* for patient *i* if and only if

$$u_{ii} \ge u_{ik}, k = 1, \ldots, J.$$

If the error terms  $\varepsilon_{ik}$  is independently and identically distributed as an extreme value,<sup>5</sup> then

Prob
$$(y_i = j) = \frac{e^{\beta_i x_i + \gamma_j z_j}}{\sum_{k=1}^{J} e^{\beta_k x_i + \gamma_k' z_k}},$$

where  $y_i$  is the drug selected by the physician. By fitting a multinomial logit model we obtain estimates of the parameters of the criterion function.

As usual, because the choice model is estimated using data from patients who were given at least one of the J possible drugs, only J-1 criterion functions can be estimated and only the differences in coefficients between the J-1 drugs and the base drug can be determined.

5. See chapter 5 of Ben-Akiva and Lerman (1985) for derivations.

### 9.3.2 Data and Empirical Strategy

The MHW collects data for the Survey of Drugs from the same source used to compile the Survey on Socialized Medicine: reimbursement claims submitted by clinics and hospitals. The drug data contain drug identification codes, quantities prescribed, and the cost of drugs to individual patients. We have limited access to the 1994, 1995, and 1996 Surveys of Drugs (MHW). The limitation is that the drug data contain only partial identification codes. The full drug identification code consists of twelve alphanumeric characters, of which we have access to only the first four. The first (and broadest) classification category uses the first three digits, and the second category uses the first four digits. If the price of a particular drug is unique within its four-digit group, we can identify the drug using its quantity and cost information, but if there are two or more drugs that have an identical list price, exact identification of individual drugs is impossible. A data set on individual drug names and full chemical components has been purchased separately from a commercial source for 1995, and additional codes for the other years have been added manually.

In the drug data, each observation contains provider characteristics, patient characteristics (including sex, age bracket, and major disease), and drug characteristics (including estimated out-of-pocket cost, profit margin, and four-digit grouping). There are typically a large number of drugs classified in any four-digit group, and several four-digit groups are contained in each three-digit group (table 9.7).<sup>6</sup> Thus, some aggregation of individual drugs is inevitable.

Using data on patients who were prescribed any drug in a given threedigit group, we estimate two specifications of our multinomial logit model. In the first model, we use the four-digit groups as our dependent variable. The different drugs within a four-digit group provide variation in profit margins and out-of-pocket costs. Thus, in this model, a particular fourdigit group may become a physician's favorite because it contains a very profitable drug. Whenever possible, we select the first four-digit group in a three-digit group as the base case because it usually has the smallest total drug costs of the four-digit groups.

In the second model, we use the quintiles of drug prices in a given threedigit group as our dependent variable. In addition to individual characteristics, provider characteristics, profit margins, and out-of-pocket costs, dummy variables for each of the four-digit groups are included as independent variables. Thus, this model tries to explain why expensive drugs

<sup>6.</sup> Our data set contains nine three-digit groups, twenty-nine four-digit groups, and 729 different drugs with a sufficient number of observations for fitting the selection models. See table 9.7 for details.

	,				
Three-Digit Group	N	Four-Dieit Group	Pharmaceutical Code	Sample Size	Weighted Sample Size
		-			-
Agents for ophthalmic					
use (131)	22,423	Mydriatics and preparations	1311	300	28,286
		Miotics and preparations	1312	448	35,734
		Local anesthetics for ophthalmic use	1313	195	14,524
		Antiseptics and astringents for ophthalmic use	1314	468	31,562
		Cortisone derivatives and preparations for ophthalmic use	1315	4,554	680,740
		Vasoconstrictors for ophthalmic use	1316	20	1,877
		Antibiotic and preparations for ophthalmic use	1317	846	103,419
		Alkaloids and preparations	1318	0	0
		Other agents for ophthalmic use	1319	15,592	1,900,333
Antihypertensives (214)	8,445	Ganglonic blocking agents; hexamethonium, etc.	2141	0	0
		Hydralazine and preparations	2142	112	20,149
		Rauwolfia and preparations	2143	83	13,858
		Angiotensin-converting enzyme inhibitors	2144	3,948	590,152
		Methyldopa and preparations	2145	110	17,228
		Alkaloids and preparations	2146	0	0
		Other antihypertensives	2149	4,192	573,011
Vasodilators (217)	10,356	Coronary vasodilators	2171	10,256	1,310,327
		Peripheral vasodilators	2172	66	8,099
		Other vasodilators	2179	1	2
Antihyperlipemia agents (218)	8,531	Linoleic acid and preparations	2181	0	0
		Lecithin and preparations	2182	2	150
		Clofibrate derivatives and preparations	2183	572	90,050
		Other antihyperlipemia agents	2189	7,957	1,530,655

1995 Survey of Drugs Summary Statistics

Table 9.7 1995 Surv

Agents for peptic ulcers (232)	18,833	Methylmethionine derivatives and preparations Glutammine and preparations Azulene and preparations Extract preparations of crude drugs H2-recepter antagonists	2321 2322 2323 2324 2324 2325	61 15 153 3,486	4,860 1,579 27,022 336 554,789
Analgesics, antipruritics, astringents, and anti- inflammatories (264)	57,061	Other agents for peptic ulcers Methyl salicylate and preparations Antihistaminic preparations for external use Ammonia preparations	2329 2641 2642 2643	15,112 2 177 1	2,013,021 280 12,628 220
		Lead compound and preparations; lead oxide, lead acetate, etc. Peppermint-gum and camphor-peppermint preparations Adrenocortical hormone preparations Mixed preparations compounded of antibiotics and adreno- corticoals hormone preparations	2644 2645 2646 2647 2640	0 15 11,334 3,435 3,435	0 1,179 1,359,440 474,236 5.642 188
Antimetabolic agents (422)	826	Mercaptopurine derivatives and preparations Methotrexate derivatives and preparations Fluorouracil derivatives and preparations Cytosine derivatives and preparations Other antimetabolic agents	4221 4222 4223 4224 4229	379 379 444	98.052 98,952 98,952 98,052
Antibiotic preparations acting mainly on gram-positive bacteria (613)	12,516	Penicillin derivatives and preparations Cephem derivatives and preparations Fosfomycin and preparations Other antibiotic preparations acting mainly on gram-positive and eram-reactive bacteria	6131 6132 6135 6135	707 11,485 232 92	177,937 3,028,090 54,405 26,994
X-ray contrast agents (721)	918	Iodine compounds and preparations Barium salts and preparations Other X-ray contrast agents	7211 7214 7219	54 63 801	8,084 10,055 170,401

Source: Authors' computations based on survey on drugs (MHW 1994-1996).

are chosen: Is it because of patient characteristics, provider characteristics, drug characteristics, or economic incentives? In order to maintain consistency with the first model, the base case has been set at the 1st quintile whenever possible.

One possible criticism of these estimations is that there is a considerable degree of arbitrariness in the Japanese pharmaceutical classification system. For example, within any three-digit group there is a large number of drugs whose fourth digits are nine, which denotes all "other agents" in the group. Typically these are drugs that could not be grouped with existing drugs when they were introduced. As a result, many of the most popular drugs in a group often belong to the "nine" category, although they may not be similar at all. Thus, the dummy variables for the four-digit groups may not be reliable, and we may be able to improve our estimation by using better drug classification systems in the selection function.

Fortunately, in an attempt to test the feasibility of a reference price system for Japan, a working group of MHW, consisting mainly of physicians and pharmacologists, has constructed more homogeneous groups of drugs for four categories: (a) peptic ulcers, (b) antihypertensives, (c) analgesics and others, and (d) antibiotics and others. In order to move from the official classification to these reference groups, we need to identify all drugs in a given group. We were unable to do this for (a), (b), and (c). However, we were able to identify most of the individual drugs in group (d) using price differences. Our data regarding the new homogeneous groupings within (d) are summarized in table 9.8. Since the first two groups have fewer than several hundred observations, we do not try to estimate their selection functions separately. We only estimate functions for the last three groups and for the antibiotic drugs group as a whole.

# 9.3.3 Results

### First Model

Judging from the pseudo- $R^2$  values, the estimated model fits the data very well, with most of the explanatory power provided by the profit margin variable. Of the twenty four-digit group selection functions, this variable was significant in nineteen. An example of the estimated selection function is shown for antihypertensives in table 9.9. The sizes of the coefficients are not intuitive, however, because of the comparative or conditional nature of the estimated selection functions.

To interpret the estimated profit margin coefficients, we perform a simulation in which all profit margins are set equal to zero without changing prices. The result of the simulation is summarized in table 9.10. Although the results vary from one three-digit group to another, in two of the groups the reduction in total drug costs exceeds 30 percent. In three additional groups this reduction exceeds 20 percent, and in two other groups it ex-

Chemical Components	Sample Size	Main Effect	Sample Size	Generic Name	Sample Size	Weighted Sample Size	Four-digit Pharmaceutical Code
Antibiotic preparations acting mainly on gram-positive	19	Cell wall synthesis obstruction	1	Penicillins Glycopeptides	10	330 0	6111 6113
bacteria		Protein synthesis obstruction	18	Tincomycin	18	7,664	6112
Antibiotic preparations acting mainly on gram-negative	349	Cell wall synthesis obstruction	117	Penicillins Cephems	25 92	2,145 15,537	6121 6129
bacteria		Cell membrane functional disturbance	5	Polypeptides	5	721	6125, 6126
		Protein synthesis obstruction Nucleic acid synthesis	8	Aminoglycosides	8	2,745	6123
		obstruction	219	Quinolones	219	39,516	6241
Antibiotic preparations acting mainly on gram-positive and	12,391	Cell wall synthesis obstruction	12,373	Penicillins 1 Penicillins 2	279 428	78,435 93.293	6131 6131
gram-negative bacteria				Cephems 1	3,942	1,024,126	6132
				Cephems 2	7,492	1,908,932	6132
				Faropenem	0	0	6139
				Fosfomycin	232	53,601	6135
		Protein synthesis obstruction	6 7	Thiamphenicol	67	3	6249
		Folacin synthesis obstruction	16	Sulta compounds	16	2,694	6212, 6213, 6219
Antibiotic preparations acting mainly on gram-positive	1,827	Protein synthesis obstruction	1,827	Macrolides 1 Macrolides 2	95 1.476	26,227 392.654	6141 6149
bacteria and mycoplasma				Macrolides 3	191	46,373	6143, 6145, 6146
				Macrolides 4	65	16,585	6149 6142
Antihiotic menarations acting	6 240	Drotain conthacis obstruction	376	Chloramphenicol		0 260	6151
mainly on gram-positive and	2 1 2			Tetracyclines 1	' =	2.991	6152
gram-negative bacteria,				Tetracyclines 2	214	55,104	6152
rickettsia, and chlamydia		Nucleic acid synthesis obstruction	6,023	Quinolones	6,023	1,248,396	6241
Source: MHW (1999).							

Antibiotics Reference Price Groups, 1998-1999

Table 9.8

	Group	2144	Group	2149
Variable	Coefficient	Z	Coefficient	Ζ
Profit margin	0.266	223.058	0.220	186.088
Out-of-pocket costs	0.004	3.851	0.010	8.927
Number of beds	0.000	8.800	0.000	7.598
Female	-0.218	-18.495	-0.171	-14.843
Age 0–19	13.699	0.083	14.766	0.090
Age 20–39	-0.440	-9.121	-0.677	-14.804
Age 40–59	0.112	5.989	0.301	16.380
Age 65–75	-0.340	-17.322	-0.023	-1.200
Age 76+	-0.207	-8.170	0.045	1.811
Circulatory diseases	0.115	9.415	0.168	14.057
Gm worker	-0.419	-18.241	0.021	0.925
Gm dependent	-0.647	-29.640	-0.722	-33.759
Gm elderly	0.245	7.486	0.193	6.082
Nh elderly	0.171	6.460	0.195	7.670
National	-0.291	-17.518	-0.115	-7.092
University	0.760	10.237	0.955	13.025
Clinic with beds	0.557	30.631	0.634	35.622
Clinic without beds	-0.707	-45.492	-0.910	-59.956
Constant	-1.571	-62.090	-0.800	-32.398
Ν	1,214,398			
Log-likelihood	-883,404.8			
LR $\chi^{2}$ (36)	270,078.53			
$Prob > \chi^2$	0.0000			
Pseudo $R^2$	0.1326			

### Table 9.9 Selection Function for Antihypertensives, Model 1 (multinomial regression)

Source: Authors' computations based on surveys on drugs (MHW 1994-1996).

*Note:* Outcome choice circulatory diseases = a dummy variable for patients with circulatory disease as their primary disease. Gm- = patients under government managed insurance. Nh= patients under town- or city-managed national health insurance. National = patients of hospitals or clinics run by the Japanese government. University = patients of teaching hospitals.

ceeds 10 percent. Overall, the reduction amounts to almost 20 percent of drug costs.

### Second Model

We estimate the selection functions for all nine groups, but we only report the results for antihypertensives (214; table 9.11). Roughly speaking, in the second model, the physician selects the cost of the drug. This straightforward model makes it very easy to interpret the results. The magnitudes of the coefficients on the profit margin variables are very stable, but their pattern of variation suggests that profit margins are relatively more important in inducing selection of higher-priced drugs. Also, by compar-

Table 9.10 Zero-Profit Simulation Results, Model 1	lation Results, <b>N</b>	1 Iodel 1								
	Thurse Disate		Munchan		D64	Total C	Total Costs (millions of yen)	ns of yen)		Zana Duc 64
Drug Groups	Group	Group	Number of Drugs	Cost per Month (yen)	Pront per Month (yen)	Actual	Estimated	Simulated	error (%)	Zero Pront Changes (%)
Agents for ophthalmic use	131	1310	4	534	68	6	6	19		
		1311	11	5,532	754	156	156	92		
		1312	6	907	121	32	32	36		
		1314	1	13	1	0	0	6		
		1315	25	503	90	343	343	315		
		1317	12	631	86	65	65	116		
		1319	64	1,691	261	3,210	3,210	2,090		
		Total	126	1,366	212	3,816	3,816	2,676	0.0	-29.9
Antihypertensives	214	2140	21	1,654	273	85	85	1,070		
		2144	23	5,523	983	3,260	3,260	695		
		2149	71	4,755	845	2,720	2,720	2,090		
		Total	115	4,997	888	6,065	6,065	3,855	0.0	-36.4
Vasodilators	217	2170	8	1,053	132	6	8	650		
		2171	63	3,378	678	4,430	4,430	739		
		Total	71	3,364	675	4,439	4,438	1,389	0.0	-68.7
Antihyperlipemia agents	218	2180	6	3,917	626	353	373	1,920		
		2189	22	5,843	807	8,940	8,910	6,600		
		Total	31	5,735	797	9,293	9,283	8,520	-0.1	-8.2
Agents for peptic ulcers	232	2320	12	463	92	16	16	49		
		2325	16	5,352	1,058	2,970	2,970	737		
		2329	63	2,031	450	4,090	4,090	4,790		
		Total	91	2,718	575	7,076	7,076	5,576	0.0	-21.2
(continued)										

Table 9.10(continued)										
	Three Divit	Three Direit Four Direit	Mincher	Cost nor	Drofft ner	Total C	Total Costs (millions of yen)	ns of yen)	л. Слад	Zaro Droft
Drug Groups	Group	Group		Month (yen)	(	Actual	Estimated	Actual Estimated Simulated	(%)	Changes (%)
Analgesics, antipruritics, astringents,	ts, 264	2640	6	471	52	7	7	858		
and anti-inflammatory agents		2646	65	1,141	234	1,550	1,550	862		
		2647	13	713	115	338	338	894		
		2649	62	1,443	342	8,140	8,140	5,280		
		Total	149	1,340	307	10,035	10,035	7,894	0.0	-21.3
Antimetabolic agents	422	4220	22	31,929	5,736	3,120	3,120	3,510		
		4229	3	43,910	6,870	4,310	4,310	3,770		
		Total	25	37,931	6,304	7,430	7,430	7,280	0.0	-2.0
Antibiotic preparations acting mainly	nly 613	6130	11	1,265	311	118	103	1,760		
on gram-positive and gram-negative	tive	6131	20	1,448	313	225	258	664		
bacteria		6132	47	1,799	427	5,450	5,450	2,590		
		Total	78	1,761	418	5,793	5,811	5,014	0.3	-13.7
X-ray contrast agents	721	7211	5	1,607	208	29	29	87		
		7214	5	20,777	4,885	3,540	3,540	2,790		
		Total	43	18,933	4,435	3,569	3,569	2,877	0.0	-19.4
Total			729			57,515	57,523	45,081	0.1	-18.8

Source: Authors' computations based on surveys on drugs (MHW 1994-1996).

		A TELIAN ING GIMME		מווזמו וטפוו וענויו	(more			
				Daily Cost Quintile	Quintile			
	-		2		3		5	
Variable	Coefficient	ы	Coefficient	ы	Coefficient	ы	Coefficient	ы
Profits per day	-1.20	-180.15	-0.35	-247.45	-0.13	-243.29	0.01	28.10
Out-of-pocket costs	-0.30	-45.38	-0.16	-88.46	-0.05	-65.40	0.05	185.48
Number of beds	0.00	25.08	0.00	23.50	0.00	30.52	0.00	21.35
Female	-0.56	-23.48	-0.26	-21.93	-0.11	-18.52	0.14	28.06
2144 dummy	-74.08	0.00	-38.50	-496.15	-34.97	-470.42	-0.74	-48.89
2149 dummy	-35.82	-500.41	-34.95	-486.25	-33.47	-449.69	-0.17	-11.30
Age 0–19	-31.12	0.00	-0.11	-0.29	1.91	5.33	-6.75	-55.19
Age 20–39	2.41	15.59	-0.24	-2.57	-1.67	-28.67	-1.31	-39.03
Age 40–59	0.50	12.16	0.14	6.76	0.05	5.30	-0.21	-27.21
Age 65–75	-0.60	-13.35	0.55	24.66	0.08	7.71	-0.25	-27.00

Selection Functions for Antihypertensives, Model 2 (multinomial logit regression)

-35.67 18.53 155.09

0.42 0.10

11.70 5.78

0.16

24.50 -30.17 97.68 -28.63 -81.66 -90.64 28.50 -11.88 30.05 -6.66 30.72

0.66 -0.39 -3.23

> -17.15 36.84 -31.38 47.08 11.03 22.73

-0.42 -2.09

Diseases of the eyes

Age 76+

Gm dependent

Gm elderly

Gm worker

1.48

0.08

70.89 -40.03 -58.14 28.97 -56.81

-1.38 -0.56

0.04

-1.65 -1.61 0.25

> -4.080.47 -0.640.52

0.11 43.77

24.41 -8.71

 $\begin{array}{c} 0.36 \\ -3.02 \\ 0.80 \end{array}$ 

University Nh elderly

National

-0.26 47.98

Clinic without beds Clinic with beds

Constant

-0.91-3.91

13.61

 $0.72 \\ -2.62 \\ -3.30$ 

23.73 147.20 151.93

 $\begin{array}{c} 1.94 \\ 0.25 \\ 2.76 \\ 0.36 \\ 0.52 \end{array}$ 

49.92 -21.05 48.67 13.98

> -0.39 0.102.88

-26.91 -6.88 -37.40

-0.84 -0.06 -0.31 38.31

542.33

Table 9.11

Source: Authors' computations based on surveys on drugs (MHW 1994-1996).

-1,029,430.21,229,246.58 1,214,298

Log-likelihood

N

LR  $\chi^{2}$  (80)  $Prob>\chi^2$ Pseudo  $R^2$ 

0.0000

0.3738

Note: Outcome of 4th quintile for nonelderly male covered by national health insurance program treated in a private hospital is the comparison group. Gm- = patients under government managed insurance. Nh- = patients under town- or city-managed national health insurance. National = patients of hospitals or clinics run by the Japanese government. University = patients of teaching hospitals. Diseases of the eyes = a dummy variable for patients of ophthalmologists. Z statistics are in parentheses. ing the dummy variables<sup>7</sup> for 2144 and 2149 we see that while 2144 is more popular, 2149 probably does not offer many products in the lower three price zones. Patient characteristics seem to be important in deciding which price zone to select: The elderly seem to be very important in accounting for the highest-priced drugs.

In order to give an idea of the interpretation of the size of the estimated profit margin coefficients, we carry out a simulation identical to that in the first model. The results of the simulation are summarized in table 9.12. In four groups the reduction in total drug costs exceeds 50 percent. In four additional groups this reduction exceeds 30 percent, and in the remaining group it exceeds 4 percent. Overall, the reduction amounts to 45 percent of drug costs.

### Using Reference Price Group Information

Using the framework of the second model and the reference group information of the Working Committee of the MHW in 1998, we estimate selection functions for three individual reference groups and antibiotics (as a whole). Presumably, drugs within reference groups are close substitutes, but there seems to be important heterogeneity among them. For example, we have noticed that in some cases the most expensive drugs are not necessarily the most profitable, even though they are used most frequently and were introduced most recently. One can think of several reasons why this may happen:

1. Newer drugs may be preferred by physicians because they reflect technological innovation and hence are better in some sense, even though they may not be the most profitable drugs in the group.

2. Manufacturers may be more reluctant to offer large discounts for new drugs because doing so will shorten their economic life.

3. The government may have become more stringent in pricing new drugs in an attempt to control drug costs.

In order to capture these effects we add a new variable indicating the year each drug was introduced. There are two exceptions to this rule: (a) if the drug is a "me-too" drug, we chose the first year the particular chemical was approved; (b) if the drug had already been approved at the beginning of our sample we treat it as having been introduced in 1967.<sup>8</sup> If physicians prefer drugs that are more profitable, "new," or both, we expect the year variable to work in the same way as the profit margin variable.

Table 9.13 reports the results for the largest group, antibiotic agents for gram-negative and –positive bacteria. The estimation results are fairly con-

<sup>7.</sup> The base-case four-digit class is 2140 in this case.

<sup>8.</sup> Nihon-seiyaku-danntai-rengoukai, Hoken-yakka-kenkyu-iinkai (1997) includes data on drugs that were approved after 1967.

Table 9.12 Zero-Profit Sim	Zero-Profit Simulation Results, Model 2	1odel 2								
	Ē	Ċ	1	Ċ		Tota	Total Costs (million yen)	llion yen)	μ	
Drug Group Description	Linee-Digit Code	Quintile	of Drugs	Costs per Month	Month	Actual	Estimated	Zero Profits	EITOT (%)	Changes (%)
Agents for ophthalmic use	131	1	26	67	23	4	44	73		
		2	26	330	57	66	66	160		
		3	26	524	82	397	397	526		
		4	24	1,064	150	640	640	506		
		5	24	3,859	602	2,640	2,640	337		
		Group total	126	1,366	212	3,820	3,820	1,602	0.0	-58.1
Antihypertensives	214	1	23	673	87	28	28	263		
		2	22	1,799	323	158	158	388		
		3	24	3,548	641	928	928	1,480		
		4	27	5,355	1,013	2,660	2,660	581		
		5	19	7,012	1,147	2,290	2,290	586		
		Group total	115	4,997	888	6,064	6,064	3,298	0.0	-45.6
Vasodilators	217	1	15	829	120	71	71	1,020		
		2	14	1,592	497	297	297	3		
		33	15	2,689	478	828	828	108		
		4	16	2,976	714	624	624	б		
		5	11	4,944	926	2,610	2,610	223		
		Group total	71	3,364	675	4,430	4,430	1,357	0.0	-69.4
Antihyperlipemia agents	218	1	9	594	78	ε	ŝ	146		
		2	9	946	114	6	6	319		
		ŝ	7	1,466	370	67	67	62		
		4	5	2,731	478	209	209	2,290		
		5	7	6,069	833	9,010	9,010	951		
		Group total	31	5,735	797	9,298	9,298	3,768	0.0	-59.5
Agents for peptic ulcers	232	1	22	426	<i>LT</i>	114	114	389		
		2	17	1,041	219	573	573	1,150		
		ŝ	18	2,005	571	626	626	178		
		4	17	2,581	608	2,470	2,470	602		
		5	17	6,397	1,156	3,290	3,290	1,650		
		Group total	91	2,718	575	7,073	7,073	3,969	0.0	-43.9
(continued)										

_	Ē	Ċ		C		Tota	Total Costs (million yen)	lion yen)	F	19 - C
Drug Group Description	Linree-Digit Code	Cost Quintile	Number of Drugs	Costs per Month	Month	Actual	Estimated	Zero Profits	error (%)	Lero Pront Changes (%)
Analgesics, antipruritics, astringents,	264	- 1	29	406	51	37	37	727		
and anti-inflammatory agents		7	31	565	120	120	120	186		
		ŝ	28	894	182	2,030	2,030	3,410		
		4	31	1,126	231	1,330	1,330	918		
		5	30	1,747	425	6,520	6,520	1,290		
		Group total	149	1,340	307	10,037	10,037	6,531	0.0	-34.9
Antimetabolic agents	422	1	5	7,897	2,203	43	43	57		
		2	5	11,864	822	89	89	1,660		
		33	4	17,998	3,559	191	191	204		
		4	9	30,388	5,554	857	857	627		
		5	5	43,354	7,091	6,240	6,240	730		
		Group total	25	37,931	6,304	7,419	7,419	3,278	0.0	-55.8
Antibiotic preparations acting mainly	613	1	16	579	157	26	26	170		
on gram-positive and gram-negative		2	15	871	210	72	72	301		
bacteria		3	15	1,389	356	446	446	243		
		4	16	1,553	418	1,220	1,220	107		
		5	16	1,960	441	4,030	4,030	4,710		
		Group total	78	1,761	418	5,793	5,793	5,531	0.0	-4.5
X-ray contrast agents	721	1	6	1,180	142	18	18	50		
		2	8	7,160	1,597	113	113	98		
		ŝ	10	12,211	2,607	290	290	1,140		
		4	8	22,670	5,366	2,050	2,050	892		
		5	8	25,221	6,004	1,100	1,100	0		
		Group total	43	18,933	4,435	3,571	3,571	2,180	0.0	-39.0
		Total				57,504	57,504	31,513	0.0	-45.2

Source: Authors' computations based on surveys on drugs (MHW 1994-1996).

negative B	Bacteria, Model 2		
Variable	Coefficient	Ζ	Marginal Effect
	Quintile 1		
Profit margin	-0.16	-212.57	-0.16
Year introduced	-0.57	-155.64	-0.88
Number of beds	0.00	-9.26	0.00
Female	-0.85	-37.84	-0.95
Age 0–19	-2.50	-46.14	-3.24
Age 20–39	0.21	3.99	0.33
Age 40–59	-0.96	-17.19	-0.95
Age 65–75	-2.21	-27.50	-2.23
Age 76+	-3.29	-25.35	-3.41
Circulatory diseases	1.66	30.23	1.70
Respiratory diseases	1.35	49.45	1.41
Gm worker	0.41	12.70	0.28
Gm dependent	-0.54	-20.11	-0.62
Gm elderly	0.23	1.71	0.22
Nh elderly	0.90	9.88	0.95
Public Hospital	0.04	0.61	0.02
University	-9.53	-0.07	-9.22
Clinic with beds	0.82	16.62	0.73
Clinic with beds	0.53	10.84	0.42
Constant	35.71	173.81	0.42
Multinomial logit regression	55.71	1/5.01	
0 0			
Ν	3,158,387		
Log-likelihood	-1,644,565.5		
LR $\chi^{2}$ (76)	3,145,966.69		
$Prob > \chi^2$	0.0000		
Pseudo $R^2$	0.4889		
	Quintile 2		
Profit margin	-0.10	-345.91	-0.10
Year introduced	-0.22	-152.29	-0.52
Number of beds	-0.01	-44.42	-0.01
Female	0.05	4.93	-0.04
Age 0–19	0.63	23.26	-0.11
Age 20–39	0.84	30.62	0.97
Age 40–59	-0.63	-20.36	-0.61
Age 65–75	-1.45	-35.47	-1.47
Age 76+	-1.27	-22.42	-1.39
Circulatory diseases	1.35	52.05	1.39
Respiratory diseases	0.76	58.66	0.82
Gm worker	-0.18	-10.78	-0.31
Gm dependent	-0.65	-51.95	-0.73
Gm elderly	0.36	6.19	0.35
Nh elderly	0.83	18.49	0.89
Public Hospital	0.27	10.31	0.26
University	-11.64	-0.07	-11.33
Clinic with beds	0.25	8.91	0.15
Clinic without beds	0.25	9.65	0.15
Constant	15.44	171.22	0.10
	10.77	1/1.44	
(continued)			

# Table 9.13Selection Function for Antibiotic Agents for Gram-positive and Gram-<br/>negative Bacteria, Model 2

Table 9.13(continue)	ed)		
Variable	Coefficient	Ζ	Marginal Effect
	Quintile 3		
Profit margin	-0.02	-169.93	-0.02
Year introduced	0.11	134.24	-0.19
Number of beds	0.00	-20.65	0.00
Female	-0.15	-28.11	-0.24
Age 0–19	2.06	126.79	1.32
Age 20–39	0.42	24.24	0.55
Age 40–59	0.24	13.29	0.25
Age 65–75	-0.04	-1.73	-0.06
Age 76+	0.26	8.05	0.14
Circulatory diseases	0.20	14.34	0.24
Respiratory diseases	0.06	8.93	0.11
Gm worker	-0.04	-3.83	-0.17
Gm dependent	-0.02	-4.01	-0.10
Gm elderly	0.06	1.72	0.06
Nh elderly	0.33	12.85	0.38
Public Hospital	0.19	14.38	0.18
University	-0.87	-10.09	-0.56
Clinic with beds	1.07	84.55	0.98
Clinic without beds	0.73	58.77	0.63
Constant	-8.70	-160.91	0102
	Quintile 4		
Profit margin	n.a.	0.00	
Year introduced	n.a.	-0.30	
Number of beds	n.a.	0.00	
Female	n.a.	-0.10	
Age 0–19	n.a.	-0.74	
Age 20–39	n.a.	0.12	
Age 40–59	n.a.	0.01	
Age 65–75	n.a.	-0.02	
Age 76+	n.a.	-0.12	
Circulatory diseases	n.a.	0.04	
Respiratory diseases	n.a.	0.05	
Gm worker	n.a.	-0.13	
Gm dependent	n.a.	-0.08	
Gm elderly	n.a.	-0.01	
Nh elderly	n.a.	0.05	
Public Hospital	n.a.	-0.01	
University	n.a.	0.31	
Clinic with beds	n.a.	-0.09	
Clinic without beds	n.a.	-0.11	
Constant	n.a.		
-			

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sistent with those from the other three groups. Larger profit margins and later introduction years increase the selection of more expensive drugs. The sensitivity is larger for more expensive drugs. In other words, physicians can be very easily persuaded to choose more expensive drugs if either their profit margins are larger or they are newer. They can also be very eas-

Variable	Coefficient	Ζ	Marginal Effect
	Quintile 5		
Profit margin	0.00	55.15	0.00
Year introduced	0.74	870.17	0.44
Number of beds	0.00	23.74	0.00
Female	0.27	68.52	0.17
Age 0–19	1.46	175.51	0.72
Age 20–39	-0.40	-47.88	-0.27
Age 40–59	-0.08	-9.66	-0.07
Age 65–75	0.06	5.18	0.04
Age 76+	0.25	15.67	0.13
Circulatory diseases	-0.14	-17.27	-0.10
Respiratory diseases	-0.15	-33.27	-0.10
Gm worker	0.34	59.08	0.21
Gm dependent	0.21	41.95	0.13
Gm elderly	0.01	0.34	0.00
Nh elderly	-0.20	-16.24	-0.15
Public Hospital	-0.01	-0.75	-0.02
University	-0.58	-24.35	-0.27
Clinic with beds	0.02	3.23	-0.07
Clinic without beds	0.12	20.42	0.02
Constant	-46.97	-824.83	

### Table 9.13(continued)

Source: Authors' computations based on surveys on drugs (MHW 1994-1996).

*Note:* Outcome of 4th quintile for nonelderly male covered by a national health insurance program treated in a private hospital is the comparison group. "Gm" stands for a person covered by health insurance managed by government. "Nh" stands for a person covered by a national health insurance program. N.a. = not applicable.

ily persuaded not to use the least expensive drugs if either their profit margins are smaller or they are older. For antibiotic drugs as a whole the same tendencies are confirmed, but much more emphasis seems to be placed on recent introduction, particularly for the most expensive drugs.

Again, we have carried out simulations for the zero profit-margin case (table 9.14). For the first group, the reduction in total costs is about 30 percent. For the second group, this reduction is about 20 percent, and for the third group, it is computed to reach close to 70 percent, although this figure may be unrealistic.

# 9.4 Inefficiencies in the Japanese Drug Price System

### 9.4.1 Inefficiency in the Production of Health Care Services

Under the present drug-pricing policy, providers of health care services are subject to strong economic incentives to prescribe as many drugs as possible to their patients. The wrong economic incentives can distort providers' decisions in two ways. First, they may influence the physician's se-

			č	Weighte	Weighted Costs (Million Yen)	n Yen)	Change i	Change in Costs (%)
	Daily Cost	Sample Size	Size			Taro	Actual to	Estimated to
	Quintiles	Unweighted	Weighted	Actual	Estimated	Profit	Estimated	Zero Profit
Gram-positive and gram-negative bacteria	1	105	24,517	12	12	194		
) )	2	295	85,731	74	74	1,360		
	ŝ	702	191,093	258	258	151		
	4	5,132	1,289,103	2,100	2,100	324		
	5	6,139	1,567,943	3,300	3,300	1,840		
	Total	12,373	3,158,387	5,744	5,744	3,869	0.0	-32.6
Gram-positive bacteria and mycoplasma	1	22	3,884	2	2	6		
	2	56	15,721	11	11	7		
	3	193	48,051	53	53	72		
	4	77	20,619	29	29	268		
	5	1,479	393,564	938	938	478		
	Total	1,827	481,839	1,032	1,032	829	0.0	-19.7
Gram-positive and gram-negative bacteria,	1	17	6,792	7	2	23		
rickettsia, and chlamydia	2	176	48,782	45	45	822		
	3	648	117,068	273	273	562		
	4	1,833	386,980	1,180	1,180	251		
	5	3,575	747,669	3,600	3,600	69		
	Total	6,249	1,307,291	5,100	5,100	1,726	0.0	-66.2
General baccilus	1	151	34,650	14	14	246		
	2	582	162,809	134	134	1,970		
	3	1,956	512,459	721	721	740		
	4	10,830	2,717,233	5,260	5,260	786		
	5	7,316	1,591,721	5,930	5,930	4,110		
	Total	20,835	5,018,872	12,059	12,059	7.852	0.0	-34.9

Source: Authors' computations based on surveys on drugs (MHW 1994-1996).

Summary of Zero Profit-Margin Simulations for Antibiotic Agents for Gram-positive and Gram-negative Bacteria

Table 9.14

lection decision among competing drugs, and second, they may influence the physician's selection decision between drugs and other factors of production, including physician services:

1. If providers are to produce their services efficiently, they must select the least-cost combination of production factors from all feasible combinations of inputs. If drug A and drug B have identical chemical compositions, the efficient provider must choose the one that costs less. But a system that gives differential profit margins across drugs tends to drive providers toward using the drug with the highest profit margin. This is usually the drug with the higher price. We have shown that this inefficiency can amount to as much as 30 percent of total expenditures on drugs.

2. In many realistic cases, moreover, there is a certain degree of substitutability between drugs and physicians' time (counseling). Under the present medical pricing system, in which physicians' time is not adequately reimbursed and profit margins are provided for drugs, it is certain that drugs are selected over other time-intensive procedures.

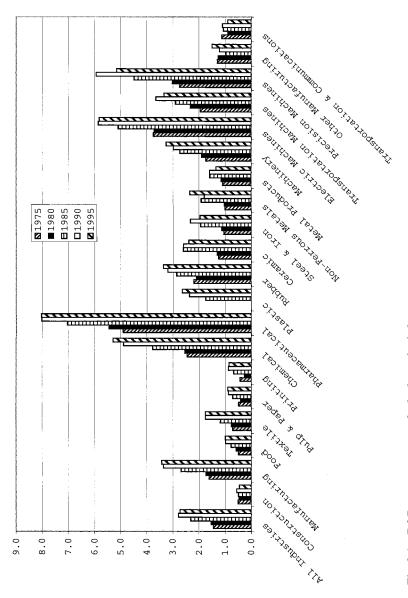
# 9.4.2 Substitution of New Drugs

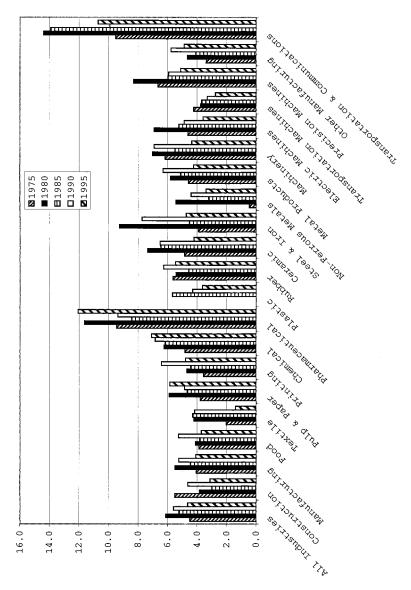
If drug discounts had remained within the reasonable zones, drug prices could not have been continuously lowered throughout the 1990s. This is evidence that competition for larger shares among drug companies has been very effective in driving down prices. We should, however, also note that large numbers of drugs have been introduced simultaneously. In fact, as we have seen in the decomposition analysis, the drug companies seem to have been able to offset reductions in prices of existing drugs by introducing new drugs and setting their prices sufficiently high.

The introduction of new drugs, however, came at a cost. The drug companies have been engaged in some of the most intensive research and development activity in Japan's major industries. In 1995, for example, the industry spent 640 billion yen out of sales of 8 trillion yen, or 8 percent, on research and development (R&D). In contrast, one of the most competitive industries in Japan, the electronics and communications industry, spent only 5 percent of sales on R&D (figure 9.1). Such an intensive R&D effort in our drug industry may have been encouraged by several factors: (a) The costs of R&D for new drugs are taken into account when the government sets the reimbursement price for the first time; (b) high prices make it possible for firms to offer discounts to providers, thereby rapidly expanding sales and profitability; and (c) without new drugs, firms' profitability declines continuously as the prices of existing drugs are repeatedly cut by the government.

### 9.4.3 Dynamic Inefficiency

The Japanese drug industry, with the second largest domestic market in the world, has enjoyed very high profitability (figure 9.2). It is supported by







	Production	All Exports	Exports Asia	All Imports	Imports from the United States	Imports from Europe
1985	4,001	30	18	308	129	162
1986	4,280	30	19	330	122	193
1987	4,825	30	19	355	117	225
1988	5,059	28	21	376	118	242
1989	5,502	27	19	417	123	278
1990	5,595	36	18	469	133	322
1991	5,697	42	24	485	129	343
1992	5,574	49	26	588	129	433
1993	5,695	44	23	584	131	430

 Table 9.15
 Japanese Drug Imports and Exports (Billion Yen)

Source: MHW (1999), Yearbook of Production Statistics for the Pharmaceutical Industry 1999

the highest R&D rates among Japan's major industries. The story, however, may not lead to such a happy ending if one looks into the industry more closely. In fact, it is possible that these impressive R&D expenditures may have actually weakened the industry rather than strengthening it. To the extent that they were induced by regulation, these R&D expenditures are simply the costs of operating under this unproductive regulation.

The statistics on the international drug trade are our first clue that there is something very strange about the industry. According to MHW statistics, drug exports amount to less than one-tenth of drug imports. Furthermore, in terms of the regional pattern of trade, almost half of the exports go to countries in Asia, whereas most of the imports are from the United States and Europe (table 9.15). This indicates the presence of substantial technological gaps among these regions, with Japan in a middle position. An analysis of drug patent data from the world's 150 largest companies (Anegawa 1996) reaches a similar conclusion: 53 percent of the world's technological capital is found in the United States, 36.6 percent in Europe, and only 10.1 percent in Japan. These findings are consistent with the hypothesis that Japanese drug price controls have generated excess returns to "new" drugs, which are similar to existing drugs, or "me-too" drugs. Most of the R&D expenditure has been allocated to copycat development rather than to truly innovative drugs.

By its nature, investment in innovative drugs involves higher risk, and under the Japanese regulatory scheme, the lower-risk investment into developing "me-too" drugs has proved just as profitable. It is not surprising that one finds that most of the R&D funds of Japanese drug firms are allocated to these easier alternatives. The process of approving and pricing "new" drugs has long been notorious for its disregard of public accountability, scientific objectivity, and global standards.<sup>9</sup> In fact, there are a large

<sup>9.</sup> A new process, called good clinical practice (GCP), was adopted in 1998 as a result of foreign pressure.

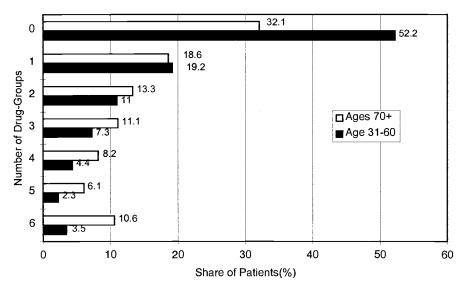


Fig. 9.3 Distribution of the number of drug groups given to outpatients with circulatory diseases

Source: Survey on Socialized Medicine (MHW 1994-1996).

number of drugs that are widely used in Japan but are not approved in the United States or Europe because their producers cannot offer sufficient evidence of increased effectiveness and superiority over existing drugs.<sup>10</sup>

# 9.4.4 External Diseconomies: Excessive Usage

The wrong economic incentives for the use of drugs may be generating very serious external diseconomies. Elderly patients are prescribed a particularly large number of drugs under the present system. According to the *Survey on Socialized Medicine* (MHW 1994–1996), the cost of drugs and injections for an elderly outpatient was 905 points<sup>11</sup> in 1996, almost twice the 477 points for the rest of the population. Using the new drug data compiled by the government,<sup>12</sup> we examine drug consumption more closely by selecting patients with circulatory diseases, the most prevalent diseases among the elderly.

Figure 9.3 illustrates that 25 percent of patients aged seventy and up received four or more groups of drugs, compared with 10.2 percent of patients aged thirty-one to sixty. Likewise, 10.6 percent of patients aged seventy and up consumed more than six groups of medicine. The risk of side effects tends to increase exponentially with the number of drugs consumed, and this is particularly acute among elderly patients with diminished liver

<sup>10.</sup> Recently, however, the MHW has changed its policy and started to remove some drugs widely used for cerebrovascular diseases due to lack of clinical effectiveness.

<sup>11.</sup> A point is 10 yen in government reimbursement schedule.

<sup>12.</sup> The drug data set is explained in detail in section 9.3.2.

and kidney functions. Moreover, many elderly patients utilize multiple providers, so the risks due to consumption of large numbers of drugs are neither controlled nor managed.

A somewhat separate issue exists with antibiotics. The distribution of antibiotics used in Japanese hospitals is concentrated very heavily among the latest generation of drugs, whereas other major countries continue to use older-generation drugs. In fact, it is not unusual to find new antibiotics being given to patients in outpatient settings with such mild afflictions as a common cold, a practice that runs a high risk of creating drug-resistant bacteria strains. This is another external diseconomy created by a regulatory system that deprives health care providers of the incentive to use less expensive, but still effective, existing drugs.

# 9.5 Various Reform Proposals

In view of the serious distortions created by the present drug-pricing policy, it is not surprising that a number of reform proposals have been advocated in the past three years. We review the merits and demerits of these proposals as reported by a government commission in January 1998.

# 9.5.1 Reimbursement of Actual Drug Costs (Liberal Democrats)

A body of ruling parties, including the Liberal Democrats, argue that the government should abandon drug price regulation altogether and move toward reimbursing providers the purchase costs of the drugs. The government commission objects to this plan on several grounds. First, patients will be charged differently from one provider to another. Second, patients will shop for better prices and choose large hospitals to get lower costs of drugs. Third, administrative costs will be huge for both insurers and providers. Fourth, providers may engage in such rent-seeking activities as utilizing extra middlemen to increase drug costs. The commission concludes that this plan should be modified at least to the extent that only the lesser of actual costs and list price is reimbursed.

# 9.5.2 Insurers Acting as Purchasing Agents

The second proposal the commission examined advocates a system in which drugs are purchased by insurers and supplied to providers as they are dispensed to their patients. Prices would be negotiated between manufacturers and insurers in the market. The commission asserts that although profit margins will vanish completely in this system, two problems will result: (a) Providers will have strong incentives to give excessive amounts of drugs to patients, as they lose track of drug costs; and (b) providers will have no incentive to use less expensive drugs. A variation on this was advocated by the Japan Medical Association (JMA) in a last-minute effort to abort the reference price system MHW was about to introduce. The JMA asked the government to purchase all drugs that providers need for patients with public health insurance.

# 9.5.3 Reference Price System of the MHW

Finding the other alternatives unacceptable, the commission considered the reference price system. In this system, first drug manufacturers announce list prices, and then the government sets a single reimbursement price for all drugs in a given homogeneous group. The reimbursement price would be set equal to a weighted average of the list prices in each group. The majority of commission members found this system to be the most desirable system, because it promotes price competition among substitutable drugs and allows manufacturers to price their own products. The member of the commission who represents JMA, however, opposed this plan strongly, presumably because JMA members would lose tremendous income by moving to this system. The JMA would agree to move to a different system only if it is allowed to recover its losses elsewhere in the fee schedule. Drug companies, particularly powerful drug companies, were not happy either. They feared that patients would not be willing to bear additional costs if they priced their products above the reimbursement prices, so they would lose part of their premium on higher-quality products. They also feared that the system would be open to considerable intervention by the MHW, as it depends critically on how broadly or how narrowly each reference group is defined. In view of this strong opposition, the ruling parties decided to mothball the plan just prior to the formation of the fiscal year (FY) 1999 government budget plan.

# 9.5.4 Removing Drug Costs from Public Health Insurance

Although no one has advocated this option so far, it is a logical alternative, given that the government will continue to regulate drug prices in one way or another as long as drugs are covered by public health insurance. It is hard to imagine, however, that a majority of Japanese will want to remove drug costs completely from the provisions of public health insurance. On the other hand, if options are offered to the public to retain drug coverage at full insurance charges or to give up the coverage for lower charges, there may be a substantial portion of the public who would be willing to take the risk or to buy private drug insurance.

# 9.6 Conclusions

We have presented empirical evidence to show that the periodic reductions in drug prices were mostly on paper and that they were largely offset by providers who switched to more expensive drugs and used a larger number of drugs for the same patients. Drug companies have continually introduced "new" drugs into the market to replace "old" drugs, and the government has approved these "new" drugs at considerably higher reimbursement rates than those of the drugs they replaced.

We have also shown that there is strong evidence that physician drug choice is influenced by drug profit margins. In fact, our results suggest that if the profit margins of drugs disappeared overnight, total drug costs would be reduced by 20 percent to 50 percent. This reduction is achieved even when the probability of using a given drug remains unchanged and drug prices are kept constant.<sup>13</sup>

In 1998, in an attempt to move away from the current price regulation system, the government commission in charge examined a number of alternative drug pricing schemes and chose the reference price system. Under this system, producers first set list prices, and then the government sets a maximum reimbursement price for what it defines as a homogeneous group of drugs. The adoption of the reference price system was blocked at the last minute by strong opposition from JMA and drug companies. Apparently the government is not yet ready to move to a free market for drugs, but any option retaining elements of regulation presents its own set of problems, some of which are fairly serious.

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13. Although these numbers are large, they are not outrageous. For instance, after the public insurance program began making lump-sum payments to health care providers for long-term elderly inpatients, the cost of injections reportedly dropped by almost 50 percent and the cost of drugs by almost 35 percent. These providers are no longer subject to economic distortions that induce them to give more drugs and shots to their patients.

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