

# Entry, Exit and the Determinants of Market Structure\*

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## Abstract

This paper estimates a dynamic, structural model of entry and exit in an oligopolistic industry and uses it to quantify the determinants of market structure and long-run firm values for two U.S. service industries, dentists and chiropractors. Entry costs faced by potential entrants, fixed costs faced by incumbent producers, and the toughness of short-run price competition are all found to be important determinants of long-run firm values, firm turnover, and market structure. Estimates for the dentist industry allow the entry cost to differ for geographic markets that were designated as Health Professional Shortage Areas (HPSA) and in which entry was subsidized. The estimated mean entry cost is 22 percent lower in these markets. Using simulations we compare entry cost versus fixed cost subsidies and find that both are similar in terms of cost per additional firm but have very different impacts on long-run profits and firm turnover.

Keywords: entry, exit, market structure, competition, service industry

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# 1 Introduction

The relationship between market structure and the competitiveness of market outcomes has played a major role in anti-trust enforcement, regulatory proceedings, and industrial organization research. While the effect of market structure, the number and relative size of producers, on firm and industry pricing, markups, and profits is generally the focus of interest, it has long been recognized that market structure cannot be viewed as exogenous to the competitive process.<sup>1</sup> Market structure is determined by the entry and exit decisions of individual producers and these are affected by expectations of future profits which, in turn, depend on the nature of competition within the market.

A two-stage model of entry and competition has provided a unifying framework for analyzing the relationship between market structure and the competitiveness of market outcomes. In the short run, the number of firms  $n$  is fixed, firms compete through price or quantity choice, and this generates profits  $\pi$  for each incumbent as a function of the market structure. This short-run relationship is captured by a function  $\pi(n)$  which Sutton (1991) terms "the toughness of competition." This function reflects product, demand, and cost factors that determine how competition occurs in the market. In the long run, the number of firms is endogenous and results from a group of potential entrants each making a decision to enter the market given knowledge of  $\pi(n)$ . The key structural equation at this stage is a zero-profit condition that guarantees that each entrant covers all fixed costs.<sup>2</sup>

Overall, this two-period framework is designed as a model of long-run market structure and it does not distinguish the continuation decision of an incumbent firm from the entry decision of a potential entrant.<sup>3</sup> If there is a difference between the fixed cost an incumbent faces and the

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<sup>1</sup>See Sutton (1991) Chapter 1, for a summary and discussion of the historical treatment of market structure in the industrial organization literature.

<sup>2</sup> Beginning with Bresnahan and Reiss (1987, 1991) empirical studies based on this two-period framework have relied on the zero-profit condition to specify a relationship between the number of firms and market size and have estimated using cross-sectional data for geographic markets. Other papers in this literature are: Berry (1992), Campbell and Hopenhayn (2005), Mazzeo (2002), Syverson (2004), and Seim (2006). Berry and Reiss (2007) discuss the conditions under which this type of data can be used to empirically distinguish the toughness of price competition from the fixed cost of entry and different sources of firm heterogeneity. Sutton (1991) uses the framework to identify a robust relationship between market size, the level of sunk entry costs, and market concentration for homogenous goods industries where the level of entry cost is determined by the production technology and is exogenous to the entrant.

<sup>3</sup>The exception to this is Berry (1992). In modeling the choice of an airline to fly between two cities, A and

sunk entry cost a potential entrant faces then these two types of firms will respond differently when market fundamentals change. In addition, without some source of firm heterogeneity, the two-period model cannot produce simultaneous flows of entering and exiting firms, something that is commonly observed in market-level data.

In this paper we estimate a dynamic structural model of firm entry and exit decisions in an oligopolistic industry and distinguish the decisions of incumbent firms from potential entrants. The model is a variant of the one developed by Pakes, Ostrovsky, and Berry (2007). We use a panel data set of small geographic markets and data on the average profits of firms and the flows of entering and exiting firms in each market to estimate three key structural determinants of entry, exit and long-run profitability. The first is the toughness of short-run price competition  $\pi(n)$ , the second is the magnitude of the sunk entry cost faced by potential entrants, and the third is the magnitude of the fixed cost faced by incumbent producers. These three components are treated as the primitives of the model, estimated, and used to measure the distinct impact of incumbents and potential entrants on long-run profitability and market structure.

We use the empirical model to analyze the entry and exit patterns of establishments in two medical-related service industries: dentists and chiropractors. Using micro data collected as part of the U.S. Census of Service Industries, we measure the number of establishments, the flows of entering and exiting establishments, and the average profit of establishments for more than 400 small geographic markets in the U.S. at five-year intervals over the 1982-2002 period. We use this data to estimate an empirical model that characterizes the determinants of long-run firm values, the entry rate, and the exit rate across markets and over time. We use the model to analyze the effect of entry subsidies that were implemented by the U.S. Health Resource and Services Administration beginning in 1978 to encourage dentists to locate in geographic markets that were designated as Health Professional Shortage Areas (HPSA). Finally, we contrast the costs and benefits of this policy with an alternative that subsidizes the fixed costs of incumbent firms.

The empirical results show that the toughness of price competition increases with  $n$ . For

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B, in a year he allows the airline's profit function to depend on whether or not they had a presence in city A or B or both in prior years. Longitudinal information, either panel data or an historical measure of market structure, are needed to distinguish the different impacts of incumbents and potential entrants on market structure and competition.

dental practices the slope of the function  $\pi(n)$  is negative, statistically significant, and particularly large as the number of establishments increases from 1 to 4. In the chiropractor industry there is also a negative effect but the decline is smaller in magnitude and not statistically significant. Estimates of the distributions of entry costs and fixed costs parameters indicate that they are statistically significant for both industries with the magnitudes being larger in the dental industry. We also find that entry costs are 22 percent lower in markets that were designated as HPSA markets for dentists. Overall, the estimates indicate that all three primitives of the model are important components of long-run firm values and market structure. As the number of firms in the market increases, the value of continuing in the market and the value of entering the market both decline, the probability of exit rises, and the probability of entry declines. These outcomes also differ substantially across markets due to differences in exogenous cost and demand factors. We use the model estimates to simulate two alternative policies that subsidize firms in underserved markets. One is an entry cost subsidy that mimics the present policy used in HPSA markets and the second is a policy that subsidizes the fixed costs of incumbent producers. The two policies impact firm decisions in very different ways. The entry cost subsidy encourages entry but lowers long-run profits and increases the exit rate from the market while the fixed cost subsidy raises long-run profits, lowers the exit rate, and only slightly raises the entry rate. A cost benefit comparison shows that both policies are similar in terms of their cost per additional firm.

The next section of this paper reviews the recent literature on structural models of entry and exit. The third section provides some background on the sources of entry and exit barriers in the dentist and chiropractor industries and summarizes the patterns of turnover observed in our data. The fourth section summarizes the theoretical model of entry and exit developed by Pakes, Ostrovsky, and Berry (2007) and presents our empirical representation of it. The fifth section summarizes our data focusing on the measurement of entry and exit, profitability, and the number of potential entrants in each geographic market. The sixth section reports the econometric estimates of the toughness of competition, entry cost, and fixed cost distributions for each industry. It also reports the results of several counterfactual exercises that reveal the importance of these three factors in generating turnover and the level of long-run profitability.

## 2 Literature Review

The theoretical and empirical literature on firm turnover has developed in parallel over the last two decades. Broad descriptions of the empirical entry and exit flows have been produced for different countries, industries, and time periods.<sup>4</sup> A common finding is that many industries are characterized by the simultaneous entry and exit of firms so that, while some producers are finding it unprofitable to remain in the industry, others are finding it profitable to enter. This leads immediately to interest in sources of heterogeneity in profits, entry costs, or fixed costs across firms in the same industry or market.<sup>5</sup> A second finding is that the level of turnover varies across industries and is correlated with the capital intensity of the industry. This leads to interest in the level of entry costs, how they act as a barrier to entry and exit, and how they vary across industries.<sup>6</sup>

Related to these empirical findings, and often building on them, is a theoretical literature that characterizes equilibrium in an industry populated by heterogeneous firms with entrants that face sunk costs of entry. The dynamic models of Jovanovic (1982), Lambson (1991), Hopenhayn (1992), Dixit and Pindyck (1994), Ericson and Pakes (1995), and Asplund and Nocke (2006) all share the feature that the participation decision for an incumbent firm differs from the decision for a potential entrant. When deciding to remain in operation, incumbents compare the expected sum of discounted future profits with the fixed costs they must cover to remain in operation, while potential entrants compare it with the sunk entry cost they must incur at the time of entry. The same future payoff can thus lead to different participation decisions by an incumbent and a potential entrant and this has implications for the way that market structure responds to changes in expected future profits. In this environment, unlike

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<sup>4</sup>For example, Dunne, Roberts, and Samuelson (1988) measure entry and exit flows for the U.S. manufacturing sector and Jarmin, Klimek, and Miranda (2008) provide similar evidence for the U.S. retail sector. Bartelsman, Haltiwanger, and Scarpetta (2008) provide a cross-country comparison of firm turnover patterns. See Caves (1998) for a summary of the earlier empirical evidence on firm turnover.

<sup>5</sup>A large empirical literature has developed that relates entry and exit patterns to underlying differences in firm productivity. For example, see Bailey, Hulten and Campbell (1992), Foster, Haltiwanger, and Krizan (2001), and Aw, Chen, and Roberts (2001).

<sup>6</sup>See Dunne and Roberts (1991) for evidence on the correlation between turnover and capital intensity in U.S. manufacturing industries. In Sutton's (1991) model for exogenous sunk cost industries, one of the key determinants of market structure is the sunk cost needed to construct a plant of minimum efficient size. This will obviously vary across industries with the capital intensity of the technology.

in the two period models of market structure, an industry's market structure at a point in time depends, not just on the expected future profit stream, but also on the past market structure. In Dunne, Klimek, Roberts, and Xu (2009) we find that market structure in the dentist and chiropractor industries depends on the lagged number of firms and the number of potential entrants in the market as implied by these dynamic theoretical models

Another insight from the theoretical literature is that the level of entry costs affects the magnitude of the flows of entry and exit. For example, in both Hopenhayn's (1992) competitive framework and Asplund and Nocke's (2006) imperfectly competitive framework, markets with high sunk entry costs are characterized by low rates of producer turnover. The sunk cost of entry acts as a barrier to entry that insulates the existing firms from competitive pressure. Industry profit and average firm value can also increase when entry costs are large.<sup>7</sup>

Recently, several authors have developed dynamic models of entry and exit in imperfectly competitive markets (Aguirregabiria and Mira (2007), Bajari, Benkard, and Levin (2007), Pendorfer and Schmidt-Dengler (2003), and Pakes, Ostrovsky, and Berry (2007)) Empirical papers that have drawn on these methodologies have utilized data on the flows of entering and exiting firms to estimate dynamic structural models of entry and exit in imperfectly competitive markets. Collard-Wexler (2006) implements a discrete choice model of entry and exit and are able to estimate parameters measuring both the toughness of competition and entry costs (where each is expressed relative to the variance of unobserved shocks to profits). Ryan (forthcoming) studies market structure in the cement kiln industry By modeling both the demand and cost curves in the industry and treating plant capacity as a dynamic choice variable, he is able to characterize markups and capital adjustment costs as well as entry costs in the industry. In this paper we exploit data on average firm profits and entry and exit flows to estimate a variant of the model of Pakes, Ostrovsky, and Berry (2007). Although we cannot estimate the level of markups we are able to measure the toughness of short run competition as well as entry costs, fixed costs, and long-run firm values in dollars.

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<sup>7</sup>There is also a selection effect which depends on the level of the entry cost. When entry costs are high, more low-profit firms will survive and this will tend to reduce industry profit and average firm value. As long as this selection effect is not too strong, the industry profitability will be positively correlated with the magnitude of sunk entry costs.

### 3 Turnover in the Market for Dentists and Chiropractors

#### 3.1 Institutional Differences in Entry and Exit Costs

In this paper we study the determinants of market structure for two health services industries, dentists (NAICS 621210) and chiropractors (NAICS 621310), that are similar in terms of the nature of demand and technology but differ in the level of profits and turnover patterns.<sup>8</sup> They both provide their services in relatively small local markets and the decision-making unit is a practice. Although there are several choices for the legal form of organization, sole proprietorship, partnership, or corporation, many of the practices are small, single doctor businesses. The market demand for these services is closely tied to population but the level of demand, and thus revenue and profits, generated by a given population will differ between the two professions. Other characteristics of the market that affect the level of total demand include income, demographics, and prevalence of insurance. This will lead to different entry flows, exit flows, and number of practitioners in the two professions. The range of products offered is fairly standardized and services of different practitioners are good substitutes for each other, at least until the population level reaches the point where specialization into different subfields (orthodontia, cosmetic dentistry) occurs. The technology is reasonably standardized across establishments in each industry and the main inputs are office space, capital equipment, office staff, and technical assistants. These are combined with the doctor's time to produce output.<sup>9</sup>

An importance difference between the two professions is the level of entry costs faced by a new practice. In our framework an entry cost is any cost born by a new establishment in a geographic market that is not born by an existing establishment. In addition to the cost of renovating office space and installing capital equipment, there is also the cost of attracting a stock of patients. Further, entry costs can arise because of entry barriers, such as state licensing restrictions, that slow the geographic mobility of dentists or chiropractors from one market to another. The entry costs vary between the two industries for a number of reasons. The simplest difference arises from the cost of capital equipment and office construction. Dental

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<sup>8</sup>Prior to 1997, dentists are SIC industry 8021 and chiropractors are SIC 8041.

<sup>9</sup>There is some room for substitution among the inputs and the importance of technician assistants, particular in dental offices, has increased over time.

offices generally require multiple treatment rooms with x-ray and dental equipment. The kind of physical infrastructure, electrical, plumbing, and support structures for x-ray equipment, tend to be very specialized and typical office space requires significant renovation to make it usable.<sup>10</sup> In contrast, the main equipment for a chiropractic office is a specialized chiropractic table in each treatment room. For both dentists and chiropractors it is possible to lease the necessary equipment which can reduce the size of the initial investment.

Another source of difference in entry conditions between the two professions involves licensing requirements.<sup>11</sup> Professionals in both fields must be licensed to practice in a state. Professional schools are typically four years in both fields, although tuition at dental schools is higher. Also, dental students typically have a bachelors degree before they enter while a significant fraction of chiropractic students do not have a bachelors degree. At the end of schooling, national written exams are given in both fields. Dentists must also pass clinical exams that are administered regionally or by individual states. The acceptance of results across states varies by state but is not uncommon. The use of regional examining boards has grown over the last 20 years and this has made it easier for new dentists to be qualified for a license in multiple states. For chiropractors there is a national exam that covers clinical skills, but some states require additional state exams.<sup>12</sup>

During the time-period of our data, there was a federal government policy in place to promote the entry of dentists in geographic markets that were designated as underserved. The designation, known as Health Professional Shortage Area (HPSA), was based on a combination of population-dentist ratio, local poverty level, and distance to other markets. The policy was

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<sup>10</sup>Osterhaus (2006) reports that the current cost of opening a new dental practice in Arizona is between \$450,000 and \$550,000. This includes the cost of construction, state-of-the-art equipment, and allowances for working capital and marketing.

<sup>11</sup>See American Dental Association (2001) and Sandefur and Coulter (1997) for further details on licensing requirements in each profession.

<sup>12</sup>Besides the licensing process by examination, by which most new graduates are licensed, there is a separate licensing process for experienced practitioners that want to relocate to a new state. Referred to as licensing by credentials, this requires a dentist to show evidence of practice experience, often five or more years of continuous practice. A gap in the practice period or disciplinary actions may disqualify an experienced dentist from obtaining a license by credentials. This can help reduce mobility of dentists. The state of Ohio reports that, of the 1046 licenses issued between 1999 and 2004, only 45 were issued by credentials. In order to increase the number of dentists in the state, the Texas legislature recently passed legislation to reduce the hurdles faced by dentists using the credentials process, specifically reducing the number of years of experience required and requiring the State Board of Dental Examiners to consider the acceptance of other regional clinical exams.

designed to subsidize the entry costs of new dentists into these markets by offering repayment of dental school loans. Repayment was based on a sliding scale ranging from \$60,000 to \$170,000 per dentist depending on the length of time they served in the HPSA. The policy was begun in 1978 and continued throughout the period of our sample 1982-2002. In our empirical model we will distinguish HPSA and non-HPSA markets and allow entry costs to differ between them.

On the exit side, we will model the shut down decision as depending on fixed costs that the firm must pay to remain in operation. Because of the differences in capital equipment and office space discussed above the fixed costs are likely to be higher for the dentist industry. One final factor that is important to recognize is that our focus is on the number of firms in operation in a market, not the identities of the doctor owning the firm. We are interested in modeling the startup and shutdown decisions of a practice that can change the number of firms in operation, not the sale of a ongoing practice that simply changes the identity of the owner. In our data, we do not treat the sale of a practice as an exit and an entry but rather as a change in ownership which does not affect market structure or profitability. To the extent possible, what we measure in the exit statistics are the number of establishments that actually shut down.

### **3.2 The Patterns of Market Structure and Market Dynamics**

In this section we summarize market structure and magnitudes of firm turnover for these two industries. The data correspond to isolated geographic markets in the U.S. which are observed at five-year intervals beginning in 1982 and ending in 2002. These markets are all relatively small, with populations that vary between 2,500 and 50,000 people. For dentists we utilize 639 geographic markets that have at least one but not more than 20 establishments. For chiropractors we use 410 markets that have between one and eight practices. The data is described in detail in Section 5 below but here we discuss the counts of establishments present in each of the years and the count of the entering and exiting establishments between each pair of years.

Table 1 aggregates the market-year observations into categories based on the number of establishments ( $n$ ) in year  $t$  and provides the mean number of entrants and exits from  $t$  to  $t+1$ .

Several patterns are important to recognize. First, as we move down the table to markets with a larger number of incumbents (which also reflects larger populations), the average entry and exit flows increase. Not surprisingly, there is more turnover in larger markets. When expressed as a proportion of  $n$ , the entry and exit patterns are more stable across market sizes. In the case of dentists, the entry proportion in Column 3 declines monotonically with  $n$  until  $n=5$ , but once the market has 5 incumbents the entry flow is between .18 and .20 with no pattern across larger markets. Similarly, with chiropractors there is an initial decline in the entry proportion as  $n$  increases, but beyond  $n=3$  there is no systematic pattern as the entry proportion varies from .29 to .41. The exit flow, expressed as a proportion of  $n$  shows little variation for dentists. With the exception of the markets with only one establishment and markets with more than 18 establishments, the exit rate lies between .17 and .21. In contrast, for the chiropractors there is a systematic increase in the exit rate as  $n$  increases. One explanation for the entry patterns and exit pattern for dentists is that we are observing markets in long-run equilibrium with relatively stable demand and cost conditions and entry and exit are determined primarily by heterogeneity in entry costs and fixed costs.

The second pattern is that the entry and exit flows, for a given level of  $n$ , are always larger for chiropractors than dentists. This holds in both absolute magnitudes and proportional to the number of firms. This suggests differences in underlying entry costs between the two industries. Finally, there is simultaneous entry and exit in many markets for both industries. The fifth column of Table 1 reports the percentage of market-year observations that have simultaneous entry and exit. The statistics indicate that simultaneous entry and exit are common, even in many markets with only a few firms. This indicates that the empirical model must recognize and allow for some form of heterogeneity in expected profitability across firms.

Overall, the entry and exit statistics suggest that a combination of competitive and technological factors interact to produce the market-level outcomes we observe and the importance of each factor differs between the two industries. In smaller markets, those with 1 to 5 firms for dentists and 1 to 3 firms for chiropractors, there is a pattern in entry and exit rates that could reflect both systematic market-level effects on profits as well as underlying firm heterogeneity. To isolate these effects we will need to estimate the profit function for producers in each indus-

try, where there is a role for both the number of firms in the market and overall market size to affect profits. The turnover statistics suggest substantial within-market turnover in both industries but a higher degree of turnover among chiropractors. One explanation for this difference is that dentists face higher sunk entry costs in establishing a business. The model developed in the next section will allow us to estimate these entry costs for each industry. Finally, the flows of simultaneous entry and exit indicate that heterogeneity exists across producers within the same market. This heterogeneity in outcomes could result from differences in fixed costs or entry costs across producers. Section 6 reports econometric results that isolate these separate effects.

## 4 A Model of Entry, Exit, and Profit

### 4.1 Theoretical Model

In this section we outline the dynamic model of entry and exit. It is very similar to the model developed by Pakes, Ostrovsky, and Berry (2007) with some modifications that aid estimation. We begin with a description of incumbent producer  $i$ 's decision to exit or remain in operation. Let  $s$  be a vector of state variables that determine the profit each firm will earn when it operates in the market. Represent the per firm profit as  $\pi(s; \theta)$  where  $\theta$  is a vector of profit function parameters. The state vector  $s = (n, z)$  contains two elements:  $n$  the number of incumbent firms in the market at the beginning of the period and  $z$  a set of exogenous profit shifters. The profit shifters in  $z$ , which will include variables that shift production costs, such as market-level input prices, and total market demand, such as market population, are assumed to evolve exogenously as a finite-state Markov process. The number of firms  $n$  will evolve endogenously as the result of the individual firm entry and exit decisions. Given a number of entrants  $e$  and exits  $x$ , the number of active firms evolves as  $n' = n + e - x$ . The individual entry and exit decisions will be determined by current and expected future profits and, through their effect on  $n'$ , will impact future profits.

In the current period with market state  $s$  each incumbent firm earns  $\pi(s; \theta)$ . At the end of the period they draw a fixed cost  $\lambda_i$  which is private information to the firm and is treated as an *iid* draw from a common cumulative distribution function  $G^\lambda$ . This fixed cost will be paid

in the next period if they choose to continue in operation.<sup>13</sup> Given the market state  $s$  and its observed fixed cost for the next period, the firm makes a decision to continue into the next period or to exit. The maximum payoff from the incumbent's current production plus discrete exit/continue decision can be expressed as:

$$V(s; \lambda_i, \theta) = \pi(s; \theta) + \max \{ \delta VC(s; \theta) - \delta \lambda_i, 0 \} \quad (1)$$

where  $VC$  is the expectation of the next period's realized value function for the firms that choose to produce. The firm will choose to exit the market if its fixed cost is larger than the expected future profits. This implies that the probability of exit by firm  $i$  is:

$$\begin{aligned} p^x(s; \theta) &= \Pr(\lambda_i > VC(s; \theta)) \\ &= 1 - G^\lambda(VC(s; \theta)). \end{aligned} \quad (2)$$

Dropping  $\theta$  to simplify the notation, the future firm value  $VC(s)$  can be defined more precisely as:

$$\begin{aligned} VC(s) &= E_{s'}^c [\pi(s') + E_{\lambda'}(\max \{ \delta VC(s') - \delta \lambda', 0 \})] \\ &= E_{s'}^c [\pi(s') + \delta(1 - p^x(s'))(VC(s') - E(\lambda' | \lambda' \leq VC(s')))] \end{aligned} \quad (3)$$

where the expectation  $E_{s'}^c$  is taken using the continuing firms' perceptions of the future values of the state variables  $s = (n, z)$ . The second line shows that, for each future state vector  $s'$ , the firm will earn the current profit  $\pi(s')$  and will produce in future periods with probability  $(1 - p^x(s'))$ . When it produces in future periods it earns the discounted expected future value net of the expected future fixed costs. This last expectation is conditional on the firm choosing to produce, so it is a truncated mean over the values of  $\lambda'$  that are less than the expected payoff from producing. This expression can be simplified if the fixed cost  $\lambda$  is distributed as an exponential random variable,  $G^\lambda = 1 - e^{-(1/\sigma)\lambda}$  with parameter  $\sigma$ . Then the truncated mean

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<sup>13</sup>The primary difference between this model and the one developed in Pakes, Ostrovsky, and Berry (2007) is that they model  $\lambda_i$  as a scrap value that the firm earns in the next period if it chooses to close. The models are similar in that they assume the realized profits of the firm in each period are composed of a common short-run payoff  $\pi(s, \theta)$  and a firm-specific component  $\lambda_i$  that is treated as an *iid* shock.

fixed cost can be written as:

$$E(\lambda'|\lambda' \leq VC(s')) = \sigma - VC(s') [p^x(s')/(1 - p^x(s'))]. \quad (4)$$

Substituting this into equation (3) the continuation value becomes:

$$VC(s) = E_{s'}^c [\pi(s') + \delta VC(s') - \delta \sigma (1 - p^x(s'))] \quad (5)$$

A market is also characterized by a pool of potential entrants where each observes the current market state  $s = (n, z)$  and makes a decision to enter at the start of the next period. Each potential entrant  $i$  also observes a private entry cost  $\kappa_i$  which is treated as an *iid* draw from a common entry cost distribution  $G^\kappa$ . This cost is interpreted as the startup cost plus fixed cost that the firm must pay if it chooses to produce in the next period. The payoff from entering depends on the evolution of the state variables  $(n, z)$  and the expectation of future state variables is taken from the perspective of a firm that chooses to enter. The expected profit payoff for a firm that chooses to enter is:

$$VE(s) = E_{s'}^e [\pi(s') + \delta VC(s') - \delta \sigma (1 - p^x(s'))] \quad (6)$$

where the expectation  $E_{s'}^e$  denotes that the expectation of future state values is from the perspective of an entering firm. The potential entrant enters if the discounted value of entry is larger than its private entry cost:  $\delta VE(s) \geq \kappa_i$ , so that the probability of entry in this model is:

$$\begin{aligned} p^e(s) &= \Pr(\kappa_i < \delta VE(s)) \\ &= G^\kappa(\delta VE(s)) \end{aligned} \quad (7)$$

To recognize the impact of the entry subsidy in HPSA markets, we will allow the entry cost distribution  $G^\kappa$  to differ between HPSA and non-HPSA markets.<sup>14</sup> Equations (2) and (7) provide the basis for an empirical model of the observed entry and exit flows in a market. To implement them it will be necessary to estimate the continuation and entry value  $VC(s)$  and  $VE(s)$ , across states and model the distributions of fixed costs and entry costs  $G^\lambda$  and  $G^\kappa$ .

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<sup>14</sup>When we allow the entry cost distribution to differ across markets we also recognize that the expectation of the future states  $E_{s'}^e$  and  $E_{s'}^c$  will differ between the markets.

Pakes, Ostrovsky and Berry (2007) show how to measure the continuation and entry values from market level data on profits, exit rates, and transition rates for the state variables. To simplify notation, define  $\boldsymbol{\pi}$ ,  $\mathbf{VC}$  and  $\mathbf{p}^x$  as vectors over the states  $(n, z)$  and define  $\mathbf{M}_c$  as a matrix giving the incumbent's perceived transition probabilities from each (row) state  $(s)$  to every other (column) state  $(s')$ . The value of continuation can be written as:

$$\mathbf{VC} = \mathbf{M}_c [\boldsymbol{\pi} + \delta \mathbf{VC} - \delta \sigma (1 - \mathbf{p}^x)]. \quad (8)$$

This equation can be solved for  $\mathbf{VC}$  as a function of  $\boldsymbol{\pi}$ ,  $\mathbf{p}^x$ , and  $\mathbf{M}_c$ :

$$\mathbf{VC} = [I - \delta \mathbf{M}_c]^{-1} \mathbf{M}_c [\boldsymbol{\pi} - \delta \sigma (1 - \mathbf{p}^x)] \quad (9)$$

Given a nonparametric estimate of  $\mathbf{M}_c$ , which can be constructed from data on the transitions patterns across states, we estimate  $\mathbf{VC}$  as a fixed point to equation (8) where  $1 - \mathbf{p}^x = G^\lambda(\mathbf{VC})$ . This method has the advantage that the probability of exit is generated consistently with the other parameters of the model but has the disadvantage of requiring that the value of continuation be solved for each state at each parameter vector. Pakes, Ostrovsky, and Berry (2007) identify an alternative method that is computationally simpler. They suggest utilizing nonparametric estimates of both  $\mathbf{M}_c$  and  $\mathbf{p}^x$  and substitute them into equation (9) to construct  $\mathbf{VC}$ . This avoids the need to resolve the value of continuation at each parameter vector. In our application we found that the solution of equation (8) was fast and that the estimates of  $\mathbf{VC}$  were very stable and chose to use the first method.

Finally the value of entry, equation (6), can also written in matrix notation. Let  $\mathbf{M}_e$  be the perceived state transition matrix from the perspective of the potential entrant then the value of entry becomes:

$$\mathbf{VE} = \mathbf{M}_e [\boldsymbol{\pi} + \delta \mathbf{VC} - \delta \sigma (1 - \mathbf{p}^x)]. \quad (10)$$

Given estimates of  $\mathbf{VC}$ ,  $\boldsymbol{\pi}$ , and  $\mathbf{M}_e$ ,  $\mathbf{VE}$  can be constructed and used with the entry condition equation (7), and entry flow data to estimate the parameters of the entry cost distribution  $G^\kappa$ .<sup>15</sup>

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<sup>15</sup>The main difference between the fixed cost model we use and the scrap value model developed by Pakes,

## 4.2 Empirical Model

The goal of the empirical model is to estimate the vector of profit function parameters  $\theta$  and parameters characterizing the distribution of fixed costs  $G^\lambda$  and entry costs  $G^\kappa$  for both the dentist and chiropractor industries. We will utilize a panel data set for a cross-section of  $m = 639$  geographic markets over  $t = 5$  time periods, for dentists and  $m = 410$  geographic markets over  $t = 5$  time periods for chiropractors. In the empirical application to each industry, for each market/year observation, the key endogenous variables are the number of establishments  $n_{mt}$ , the number of entering firms  $e_{mt}$ , the number of exiting firms  $x_{mt}$ , and a measure of the number of potential entering firms  $p_{mt}$ . Three exogenous state variables, the level of population  $pop_{mt}$ , the average real wage paid to employees in the industry  $w_{mt}$ , and real per-capita income,  $inc_{mt}$  are included as demand and cost shifters in these health care industries. To simplify the discussion below we will often combine these three exogenous variables into the state vector  $z_{mt} = \{pop_{mt}, w_{mt}, inc_{mt}\}$ .

### 4.2.1 Profit Function

Since we observe average market-level profits in our data, we are able to recover the parameters of the profit function  $\theta$ . We specify a profit function that is very flexible with respect to the number of firms, population, wage rate, and income. We assume that the average profit function for all dentist practices in market  $m$ , year  $t$  can be written as:

$$\begin{aligned} \pi_{mt} = & \theta_0 + \sum_{k=1}^5 \theta_k I(n_{mt} = k) + \theta_6 n_{mt} + \theta_7 n_{mt}^2 + & (11) \\ & \theta_8 pop_{mt} + \theta_9 pop_{mt}^2 + \theta_{10} w_{mt} + \theta_{11} w_{mt}^2 + \\ & \theta_{12} inc_{mt} + \theta_{13} inc_{mt}^2 + \theta_{14} (pop_{mt} * w_{mt}) + \\ & \theta_{15} (pop_{mt} * inc_{mt}) + \theta_{16} (inc_{mt} * w_{mt}) + f_m + \varepsilon_{mt} \end{aligned}$$

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Ostrovsky, and Berry (2007) is that the last term  $-\delta\sigma(1 - \mathbf{p}^x)$  in both equations (9) and (10) would be replaced by  $+\delta\sigma_s \mathbf{p}^x$  where  $\sigma_s$  is the parameter of the exponential distribution of scrap values. An increase in the mean scrap value will raise  $VC$  and  $VE$ , while an increase in the mean fixed cost will lower them. An higher value of  $VE$  will lead to a higher estimate of the sunk entry cost. We found that in estimating the scrap value model the estimated entry costs were higher than were reasonable given some indirect evidence we were able to construct on entry costs. We instead chose to develop the model treating the *iid* profitability shock as a fixed cost.

We include a set of dummy variables  $I(n_{mt} = k)$  to distinguish markets with  $k = 1, 2, 3, 4, 5$  establishments and would expect the per-establishment profits to decline with discrete increases in  $n$ . We also include linear and quadratic terms in  $n$  to allow the possibility of a diminishing effect of  $n$  on average profits as the number of firms increases beyond 5. For the chiropractor industry the maximum number of establishments we observe is  $n=8$ , so we simply replace  $n_{mt}$  and  $n_{mt}^2$  with two additional dummy variables to distinguish markets with 6 or 7 establishments. To control for the three exogenous state variables we include a quadratic specification in  $pop$ ,  $w$ , and  $inc$ .

Despite controlling for these state variables, it is likely that there are unobserved factors that lead to persistent differences in the level of profits across markets. This could include factors like education differences that could affect the demand for these services, the type of employers in the area, which could lead to differences in the degree of insurance coverage for health-related services, and differences in the availability of substitute products in the same or adjacent geographic markets. To control for potential profit differences across markets arising from these factors we include a market fixed effect  $f_m$  in the profit function specification. If there are persistent factors that cause differences in profits across markets and we fail to control for them, we expect the coefficients related to the number of firms  $\theta_1, \dots, \theta_7$  to be biased toward zero. That is, we will underestimate the competitive (negative) effect of an increase in the number of firms on producer profits. Finally, all other variation is captured with an idiosyncratic shock  $\varepsilon_{mt}$  that is assumed to be *iid* across markets and time. The inclusion of  $f_m$  in the profit function complicates the dynamic aspects of the model because  $f_m$  must now be treated as a state variable in the empirical model of entry and exit. We discuss treatment of this in the next section.<sup>16</sup>

Given the assumptions of the theoretical model, the number of firms is uncorrelated with the idiosyncratic shock  $\varepsilon$  and equation (11) can be estimated with the fixed effects estimator. The key assumption is that all sources of serial correlation in profits have been controlled for with the time-varying state variables, number of firms, and market fixed effect so the idiosyncratic shock does not contain any serially-correlated components that the firms use in making entry or

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<sup>16</sup>Akerberg, Benkard, Berry, and Pakes (2007) discuss this as one way to correct for serial correlation in the market-level profit data that arises from unobserved market-specific factors.

exit decisions. If  $\varepsilon$  does contain them, then it should be treated as an additional, unobserved, state variable in the model, substantially complicating the specification of the dynamic decisions.<sup>17</sup> While the profit function parameters could still be consistently estimated if there were instrumental variables available that were correlated with the number of firms  $n$  but not the idiosyncratic shock  $\varepsilon$ , it is difficult to identify good candidates for instruments. In particular, the lagged number of firms in the market  $n_{mt-1}$  is not an appropriate instrument because the combination of the dynamic decision process generating  $n$  and the serial correlation in  $\varepsilon$  means that  $n_{mt-1}$  will be correlated with  $\varepsilon_{mt}$ .<sup>18</sup>

#### 4.2.2 State Variable Transitions and the Probability of Exit

The second step of the estimation method is to estimate the two transition matrices  $M_c$  and  $M_e$  which can then be used to estimate  $VC$  and  $VE$  for each state using equations (8) and (10). Pakes, Ostrovsky, and Berry (2007) propose to estimate these objects nonparametrically by discretizing the values of the state variables and calculating the transition frequencies from the market-level panel data for each discrete state. In our case, the number of firms  $n$  is already a discrete variable. We construct a single continuous variable measuring the combined effect of the exogenous variables in  $z_{mt} = \{pop_{mt}, w_{mt}, inc_{mt}\}$  using the estimates of the profit function from stage 1. After estimating the profit function parameter vector  $\hat{\theta}$  we define a single exogenous aggregated state variable:

$$\begin{aligned} \hat{z}_{mt} = & \hat{\theta}_8 pop_{mt} + \hat{\theta}_9 pop_{mt}^2 + \hat{\theta}_{10} w_{mt} + \hat{\theta}_{11} w_{mt}^2 + \hat{\theta}_{12} inc_{mt} + \\ & \hat{\theta}_{13} inc_{mt}^2 + \hat{\theta}_{14} (pop_{mt} * w_{mt}) + \hat{\theta}_{15} (pop_{mt} * inc_{mt}) + \hat{\theta}_{16} (inc_{mt} * w_{mt}) \end{aligned} \quad (12)$$

that captures the combined contribution of income, population, and wages to profits. We then discretize the values of  $\hat{z}_{mt}$  into a small number of categories and use the mean of each category as the discrete set of points for evaluation. Denote these points as  $z_d$ . While the market fixed

<sup>17</sup>Das, Roberts, and Tybout (2007) estimate a dynamic entry model for a monopolistically competitive industry in which profit shocks are treated as serially-correlated state variables that are unobserved by the econometrician.

<sup>18</sup>Lagged values of the exogenous state variables  $z$  are candidates for instruments. We have estimated the profit function model using them as instruments but find that they are not highly correlated with the number of firms after controlling for current values of the exogenous values of the state variables. Given the complications arising from treating  $\varepsilon$  as an unobserved state variable we have chosen to limit the exogenous state variables to  $z$  and  $f_m$ .

effects are discrete, there is one for each of the geographic markets in our data set, 639 for dentists and 410 for chiropractors, and this quickly exhausts the data available. To simplify this we further classify the markets into a small number of categories based on their estimated  $\hat{f}_m$ . Denote these points as  $f_d$ .

The size of the estimated  $M_c$  and  $M_e$  transition matrices depends on the number of discrete categories in  $n$ ,  $z_d$ , and  $f_d$ . The number of discrete states is  $n_{\max} \cdot z_d \cdot f_d$ , where  $n_{\max}$  is the largest number of firms observed in any market, and the number of cells in the transition matrices are  $(n \cdot z_d \cdot f_d)^2$ . Given that in our data set  $n_{\max}$  is 20 for dentists and 8 for chiropractors, the number of cells exceeds the number of market observations even for small values of  $z_d \cdot f_d$ . To make the nonparametric estimation of  $M_c$  and  $M_e$  tractable we use 10 discrete categories for the exogenous state variable  $z_d$  and 3 categories for  $f_d$ . To reduce the dimensionality of the transition matrices further we exploit the fact that the state variables in  $z$  evolve exogenously and that the market fixed effect does not change over time, so that the transition probability used by continuing firms is:  $M_c(n', z'_d, f_d | n, z_d, f_d) = M_{nc}(n' | n, z_d, f_d) \cdot M_z(z'_d | z_d) \cdot I_{f_d}$ . Each of these smaller matrices can be estimated separately. A similar expression for  $M_e$  can be written as  $M_e = M_{ne}(n' | n, z_d, f_d) \cdot M_z(z'_d | z_d) \cdot I_{f_d}$ .<sup>19</sup>

To estimate these transition matrices, define the set of market-year observations observed in the discrete state  $(n, z_d, f_d)$  as  $T(n, z_d, f_d) = \{mt : (n_{mt}, z_{mt}, f_m) = (n, z_d, f_d)\}$ . The transition rate among states that is perceived by continuing incumbent firms in a market beginning in state  $(n, z_d, f_d)$  contains the matrix  $M_{nc}(n' | n, z_d, f_d)$  which is estimated as:

$$\hat{M}_{nc}(n' | n, z_d, f_d) = \frac{\sum_{mt \in T(n, z_d, f_d)} (n - x_{mt}) I [n_{mt+1} = n']}{\sum_{mt \in T(n, z_d, f_d)} (n - x_{mt})} \quad (13)$$

In this case  $I$  is a dummy variable equal to one if the period  $t + 1$  state is  $n'$ . This equation describes an incumbent's probability of transiting from state  $(n, z_d, f_d)$  to state  $(n', z_d, f_d)$ , conditional on not exiting.

The transition rate among states that is perceived by entering firms in a market beginning in state  $(n, z_d, f_d)$  depends on  $M_{ne}$  which is estimated as:

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<sup>19</sup>When we allow separate entry cost distributions for HPSA and non-HPSA markets we estimate

$$\hat{M}_{ne}(n'|n, z_d, f_d) = \frac{\sum_{mt \in T(n, z_d, f_d)} (e_{mt}) I[n_{mt+1} = n']}{\sum_{mt \in T(n, z_d, f_d)} (e_{mt})} \quad (14)$$

This describes a potential entrant's probability of transiting from state  $(n, z_d, f_d)$  to state  $(n', z_d, f_d)$ , conditional on entering in state  $(n, z_d, f_d)$ .<sup>20</sup>

Finally, the transition pattern for the exogenous state variables in  $z$  is estimated as:

$$\hat{M}_z(z'_d|z_d) = \frac{\sum_{mt \in T(z_d)} I[(z_{mt+1}) = z'_d]}{\sum_{mt \in T(z_d)} I[(z_{mt}) = z_d]} \quad (15)$$

The estimators in equations (13), (14), and (15) allow us to construct estimates of  $M_c$  and  $M_e$  which are components of the value of continuing or entering the market.

### 4.2.3 Fixed Costs and Entry Costs

The final stage of the estimation method focuses on the parameters of the fixed cost and entry cost distributions using the data on entry and exit flows in the market. For market  $m$  at time  $t$ , each of the  $n_{mt}$  incumbent firms makes a decision to continue or exit based on its private fixed cost and the value of continuing. Using the estimates from the first two stages, an estimate of  $VC(n, z_d, f_d)$  can be constructed for each state up to the parameter  $\sigma$  which characterizes the fixed cost distribution  $G^\lambda$ . For each market observation  $mt$ , the value of continuing is constructed from equation (8) and denoted  $\hat{V}C_{mt}(\sigma)$  to indicate that it depends on the parameter  $\sigma$ . Similarly, each of the  $p_{mt}$  potential entrants makes a decision to enter or stay out based on its private entry cost, and the value of entering. Denote this as  $\hat{V}E_{mt}(\sigma)$  to also indicate it depends on the fixed cost parameter  $\sigma$ . Denoting  $G^\kappa(\alpha)$  and  $G^\lambda(\sigma)$  as the cdf's of the entry cost and fixed cost, respectively, then the log of the probability of observing a market with  $x_{mt}$  exits and  $e_{mt}$  entrants is given by:

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<sup>20</sup>When we allow the entry cost distribution to differ for HPSA and non-HPSA markets, we estimate  $\hat{M}_{nc}$  and  $\hat{M}_{ne}$  separately for each group of markets.

$$\begin{aligned}
l(x_{mt}, e_{mt}; \sigma, \alpha) = & \tag{16} \\
& (n_{mt} - x_{mt}) \log(G^\lambda(\hat{V}C_{mt}(\sigma); \sigma)) + (x_{mt}) \log(1 - G^\lambda(\hat{V}C_{mt}(\sigma); \sigma)) \\
& (e_{mt}) \log(G^\kappa(\hat{V}E_{mt}(\sigma); \alpha)) + (p_{mt} - e_{mt}) \log(1 - G^\kappa(\hat{V}E_{mt}(\sigma); \alpha))
\end{aligned}$$

The log-likelihood for the entry and exit observations is

$$L(\sigma, \alpha) = \sum_m \sum_t l(x_{mt}, e_{mt}; \sigma, \alpha). \tag{17}$$

To implement this, we need to make assumptions about the cdf's for the entry cost and fixed cost distribution. Consistent with the theoretical model in the last section, we assume that the firm fixed cost  $\lambda$  is distributed as an exponential random variable with parameter  $\sigma$ , which is the mean fixed cost.<sup>21</sup> For the distribution of firm entry costs,  $G^\kappa(\alpha)$ , we have more flexibility to specify the shape of the distribution and will estimate the model under two different distributional assumptions. One is that it follows a chi-square distribution and the second is that it follows an exponential distribution. In each case, there is a single parameter  $\alpha$  to estimate and this parameter is the unconditional mean of the entry cost distribution.

## 5 Data

### 5.1 Definition of the Market

To estimate the model the data set must contain information on the entry flows, exit flows, average firm profits, exogenous profits shifters (*pop*, *inc*, and *w*), number of firms, and potential entrants across multiple markets. The data we use in this analysis come from US Census Bureau's Longitudinal Business Database (LBD) and Census of Service Industries. The LBD contains panel data on the identity of all employers in the United States for each year from 1977 through 2002, while the Census of Service Industries contains detailed information on revenues, costs, and geographic location for each establishment in the service sectors for the years 1977,

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<sup>21</sup>It is possible to extend this framework to allow heterogeneity in the fixed cost distribution across markets and time by modeling  $\sigma$  as function of some observable market-year characteristic. The empirical difficulty is that this new characteristic must be treated as another state variable in addition to  $n$  and  $z$ . We will report some results of this extension below.

1982, 1987, 1992, 1997, and 2002. Similar to the approach taken by Bresnahan and Reiss (1991, 1994), we focus on relatively isolated geographic markets that are away from large population centers. We are able to construct the necessary data for more than 700 incorporated census places, which are basically small to mid-sized towns and cities in rural or semi-rural areas. The markets have populations that vary from 2,534 to 49,750 people, which are larger than the range of market sizes studied by Bresnahan and Reiss. Of these markets 639 had at least one dental practice in every year and never had more than 20 practices. For the chiropractors, we limit the analysis to 410 geographic areas that had between 1 and 8 practices in every year.<sup>22</sup>

## 5.2 Measuring Entry and Exit

As discussed in Jarmin and Miranda (2002), the LBD uses both Census Bureau establishment-level identification numbers and name and address matching algorithms to track continuing establishments over time. An entrant in a market is defined as an establishment that is not present in the market in period  $t$  but is producing in the market in period  $t+5$  (the next Census year). Similarly, an exit is defined as an establishment that is in a geographic market in period  $t$  and is not in that market in period  $t+5$ . For each market, we construct the numbers of entering, exiting, and continuing establishments.

It is important to emphasize that we would like to eliminate the sale of an ongoing practice from the entry and exit statistics and have done this to the extent possible. This is in keeping with the assumptions of the model, which views the number of independent decision makers ( $n$ ) as the endogenous state variable affecting profits and the entry and exit decision as reflecting a change in the number of decision makers. In practice, however, the LBD is constructed based on following establishments at a specific location over time, but some of the linking relies on matching the name and address of the establishment across years. If the sale of a practice results in a name change, then it may not be recognized as an ongoing establishment and this will lead to an upward bias in the entry and exit rates we construct.<sup>23</sup>

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<sup>22</sup>There were very few markets which met our population criteria and had more than 20 dentist or 8 chiropractor practices.

<sup>23</sup>In longitudinal Census data, errors in the linkage for an establishment over time will appear as a simultaneous entry and exit. We have compared our entry rates for dental practices with independent information on new licenses reported by the licensing boards in several states. While not comprehensive, in cases where we can make comparisons with the census markets, it suggests that the rate of entry we measure is approximately 5

### 5.3 Market Level Demand and Cost Variables

In the profit function we include three exogenous state variables to capture differences in the evolution of profits across markets. To control for demand differences we include the population and the real per-capita income of the geographic market. Population estimates for incorporated places in each sample year are constructed from data collected in the Census Bureau's Population Estimates Program and are augmented by interpolations from the decennial population censuses for the earlier sample years. Real per-capita income is constructed at the county level using data from the Bureau of Economic Analysis and deflated by the CPI. To control for cost differences we measure the average real wage paid to employees in health services industries in the area. This is then deflated by the national CPI. Because we do not use local price deflators, variation in the wage variable will also reflect price-level differences across geographic markets, which is likely to be important in the cross-section dimension of the data.

### 5.4 Measuring Establishment Profits

The empirical model requires a measure of the average profits earned by establishments in each geographic market and time period. The relevant measure of profits in this industry is the net income earned by the dentist or chiropractor from operating the establishment. To construct this measure, we use information on revenue, payroll, and legal form of organization from the Census LBD and information on other business expenses from the American Dental Association (ADA) and the Census Bureau's Business Expenses Survey (BES). These expenses include licensing fees, costs of supplies and materials, insurance, rent, depreciation charges on capital equipment, and purchased services among other things. They capture market level differences in variable and some fixed costs.<sup>24</sup> These data sources report that expenses other than payroll are approximately 35% of a dentist's office revenues. For the offices of chiropractors, we rely

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percentage points higher (20 percent versus 15 percent, on average), but the cross-state patterns are similar and the 5 percentage point differential is similar across states. The higher rate in the census data could reflect errors in following existing practices over time or the movement of dentists into new geographic markets. While these two data sources have different units of measurement, establishments versus individuals, it is encouraging that the cross-sectional ranking of high and low entry states is similar.

<sup>24</sup>As developed in the theoretical model section, we will also incorporate a firm-specific fixed cost shock which will generate profit heterogeneity across firms within each market.

on aggregate data from the BES for industry 804 (Offices of Other Health Practitioners) that contains chiropractors. Based on the BES data, we estimate that other expenses account for 37% of a chiropractor’s office revenues.

In constructing a measure of profit, two other important features of the industries must be accounted for. First, the tax status of a firm will affect how key data items are reported. For sole proprietors and partnerships, the owner receives compensation as net income and not as payroll. For these legal forms of organization (LFO), firm pre-tax profits (net income) are revenue minus payroll minus estimated expenses. For professional service organizations (corporations), the owning dentist(s)/chiropractors are typically paid part of their compensation as a component of payroll. We use aggregate tax data to measure the share of payroll going to the owners of incorporated firms in each of these industries and adjust payroll and profits of corporation to reflect this. The second correction deals with the fact that the number of owner-practitioners will vary across medical offices and thus the level of firm profits will vary with the number of owner practitioners.<sup>25</sup> In order to make our profits comparable across offices of different scale, we normalize the profits per office by the average number of practitioner-owners across the LFO types. Thus, our final measure of profit is the net income per owner-practitioner.<sup>26</sup>

## 5.5 Measuring the Number of Potential Entrants

The empirical model requires that we measure the pool of potential entrants in each geographic market. One option that has been used in the literature is to assume that there is a fixed number of potential entrants in every market and time period. This is not realistic given the large variation in the population and number of firms we observe in our market-level data. Instead, we adopt two definitions of the entry pool that will allow it to vary with the size of each market. The first definition sets the number of potential entrants into a geographic market

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<sup>25</sup>Based on 1997 dentist data, for sole proprietors the ratio of the number of owners to offices is one to one; for partnerships there are roughly 1.8 owner-dentists per partnership; and for professional service organizations there are roughly 1.35 dentists per practice.

<sup>26</sup>A final modification is made to the profit figures to standardize the profit flow with the entry and exit flows. Using the Census data we have measured the flow of profits in census year  $t$  while the entry and exit numbers represent flows over the 5 year period between censuses. We convert the annual profits to the discounted sum over the five-year interval by  $\Pi_{mt} = \sum_{j=0}^4 \delta^j \pi_{mt}$  and with  $\delta = .95$ . In effect we treat the practice as making the decision to exit or enter based on the discounted sum of the five-year flow of profits. In addition, the discount rate used to construct  $VC$  and  $VE$  in equations (9) and (10) is the value at the end of the five year interval,  $.95^5 = .773$ .

in a time-period equal to the maximum number of different establishments that appear in the market over time minus the number of establishments already in operation. The rationale behind this definition is that in each geographic market we observe all potential entrants being active at some point in time. In each time period the pool of potential entrants is the set of establishments that are not currently active. We will refer to this as the "internal" entry pool because it is constructed using only data that is present in the Census LBD. It will also tend to covary positively with the population of the geographic market and the actual number of entering firms, resulting in an entry rate that is roughly constant across market sizes. The disadvantage of this measure is that it is affected by the overall growth in market size and the number of establishments over time. Since the number of establishments has increased over time due to exogenous growth in population, this measure is likely to overestimate the number of potential entrants, and thus underestimate the entry rate, in the early years of the sample.

This internal entry pool definition misses the fact that one of the main sources of entry into these professions is a doctor that breaks away from an existing practice to start a new practice in an area.<sup>27</sup> To capture this feature of the potential entry pool, we exploit additional data from the ADA, Federation of Chiropractic Licensing Boards (FCLB) and Bureau of Health Professionals (BHP) to estimate the number of non-owner practitioners in an area. Specifically, we measure the number of dentists that exceed the number of dental offices in the county in which each of the geographic markets is located and in the counties that are contiguous to this county. We use this number as our estimate of the pool of potential entrants for a market. We will refer to this as the "external" entry pool definition. In the case of chiropractors, we use much cruder information from the FCLB and BHP on the ratio of the number of licensed chiropractors to the number of chiropractors' offices to construct the excess pool of entrants available to start new businesses. This correction does not vary across geographic markets. We also adjust the chiropractor pool for new graduates, since they are a more important source of new entrants than in the case of dentists.

The potential entry pools are summarized in Table 2. The table reports the average

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<sup>27</sup>Industry sources (Weaver, Haden and Valachovic, (2001)) explain that most entry comes from dentists leaving an existing practice to start a new one and that few dental school graduates start new practices on their own right after school.

number of potential entrants across all observations with a given number of establishments. In all cases the number of potential entrants rises with the size of the market. For both industries, the "internal" entry pool gives a number of potential entrants that is slightly larger than the number of establishments in the market. This is also true of the "external" entry pool for the chiropractors. The main difference is between the internal and external pools for the dentist industry. In general, since there can be many adjoining counties for each market, we identify a fairly large number of dentists in those surrounding areas and it is the size of the dentist pool in these surrounding areas that determines the number of potential entrants. In general, this external entry pool will increase with the size of the geographic market but it is not as closely tied to the number of practices in the market as the internal entry pool. The difference in the number of potential entrants between the two definitions will likely affect the estimated sunk entry cost, with the larger entry pool implying a lower entry rate and correspondingly higher estimated entry costs. We will discuss the impact of this definition on the estimated parameters in the next section

## **5.6 Identifying HPSA Markets for Dentists**

Beginning in 1978 the U.S. Health Resource and Services Administration began to identify geographic areas that were underserved by dentists. These are denoted as Health Professional Shortage Areas (HPSA) based on the number of practitioners, population and income. These geographic areas vary in size from multiple census tracts to counties and are not identical to the economic census places that we use to define our geographic markets but we are able to match the HPSA areas to our geographic markets. If all or part of one of our geographic markets falls into the coverage area of an HPSA then we designate that geographic market as an HPSA market. If the HPSA designation is effective in attracting new dental practices then our observed geographic market should have a higher than expected number of practices, even if only part of the geographic market is in the HPSA area. Overall there are 59 of the 639 geographic markets in our data that are designated as HPSA areas for at least one of the years in our data. The HPSA designation does change over time for some of our geographic markets and we recognize this in the empirical model by allowing the entry cost distribution to differ

over time with the designation. However, we treat the designation at any point in time as exogenous to the firms in the market and do not attempt to model expected changes in the designation over time.

## 6 Empirical Results

### 6.1 Estimates of the Profit Function

The profit function parameters  $\theta$  are estimated both with and without market fixed effects and are reported in Table 3. The first column reports estimates for the dentist industry without the market fixed effect. If there are persistent unobserved profit determinants across markets, then in this specification the coefficients on  $n$  will be biased toward zero so that we underestimate the toughness of competition. When market fixed effects are ignored in the profit function, there is not a strong pattern of decline when the number of firms increases. The dummy variable coefficients for markets with one to four firms are positive and decline slightly as  $n$  increases but are not significant, although the coefficient for markets with  $n = 5$  actually increases and is significant, while the coefficients on  $n$  and  $n^2$  are small and not significant. In contrast, after controlling for market fixed effects, the same coefficients in column two indicate a more substantial decline in average profits with an increase in the number of firms. The dummy variable coefficients are larger in magnitude, but still not statistically significant, while the coefficients on  $n$  and  $n^2$  do indicate a significant negative competitive effect from an increase in the number of firms. The function representing the toughness of competition is summarized in Figure 1 for each of the two sets of parameter estimates. This plots the fitted value from the profit function regressions against  $n$ , holding other variables fixed at the sample means. The steeper line for the dentist industry shows the function when market fixed effects are controlled for and the attenuation bias that occurs when the fixed effects are eliminated is obvious. The slope of this function summarizes the impact of market structure on market performance in the short-run. When market fixed effects are incorporated, the mean predicted firm profit drops from 78.7 thousand dollars for a monopoly to 64.8 thousand for a duopoly to 46.8 thousand for a market with five firms, a 40.5 percent decline from monopoly markets to markets with 5 firms.

In the profit function with fixed effects, several of the other state variables have statistically significant coefficients. When evaluated at the sample means of the variables, the marginal effect of each of the three variables, population, income, and wages, are all positive. The first two effects are consistent with demand increases as market size and income increase. The wage effect is counter to what is expected if it is a cost shifter and is likely capturing an effect of cost-of-living differences across geographic markets. The most substantial effect comes from changes in income and, when expressed as an elasticity, the impact of an increase in income on profits is 1.53. This can reflect both increased use of dental services and use of more advanced services that are likely to have higher profit margins in higher-income areas.

The profit function parameters for the chiropractor industry are reported in column 3 (without market fixed effect) and column 4 (with fixed effect) of Table 3. Because the number of firms in our sample varies from one to eight across markets, we use a full set of dummy variables to model the effect of  $n$  on average profit, with the base group being the markets with eight firms. The magnitude of the fixed effects estimates indicate that an increase in  $n$  reduces average profits but the individual coefficients are not statistically significant. When evaluated at the mean of the other state variables, a market with one firm will have average annual profits of 59.7 thousand dollars, a duopoly will have 57.7 thousand, and a market with five firms 53.0 thousand, an overall decline of 11.2 percent as the market moves from monopoly to five firms. The estimated functions summarizing the toughness of competition are also graphed in Figure 1 and the attenuation bias in the estimates when we do not control for the market fixed effect is also present. The effect of the other three state variables are all positive when evaluated at the means. Interestingly, the profit elasticity with respect to income is .591, which is less than in the dentist industry. Demand and profits increase as market income rises but the lower elasticity might reflect substitution into other forms of medical care as income rises.

Comparing the estimated toughness of competition function between the two industries we see that, while it declines with  $n$  for both, it is much steeper for dentistry. This will reflect the nature of short-run competition among firms and the demand and cost characteristics that determine short-run profitability. What the estimated patterns suggest is that the actual entry of an additional firm will have a larger adverse impact on current firm profits in the dentist

industry. Among other things, this could reflect the availability of substitute products outside of the industry. If alternatives for dental care are more limited than for chiropractor services, then existing firms enjoy a higher degree of short-run market power which is then reduced as the number of providers increases.

## 6.2 Fixed Costs, Firm Values, and the Probability of Exit

The parameters of the fixed cost and sunk entry cost distributions,  $\sigma$  and  $\alpha$ , were estimated with maximum likelihood using the likelihood function for the entry and exit rates in equation (17). Each of these parameters is the mean of the underlying cost distribution the firms face, expressed in millions of 1983 dollars. Since the entry, exit, and profit flow data used in the likelihood function are measured over five-year intervals, the parameters are the costs of operating over a five-year period. Table 4 reports parameter estimates for several specifications of the model including two assumptions about the shape of the entry cost distribution (chi-square and exponential) and two assumptions about the pool of potential entrants (the internal and external pools as defined in section 5.5).

Panel A of Table 4 reports parameter estimates for the dentist industry. The estimate of the mean fixed cost varies from .369 to .373 million dollars across different specifications on the entry pool and entry cost distribution.<sup>28</sup> Obviously, this parameter estimate is completely insensitive to alternative assumptions on the entry dimensions.<sup>29</sup> In Panel B of Table 4 we report separate estimates the fixed cost parameter for both HPSA and non-HPSA markets. The point estimates in all cases are virtually identical to the estimates in Panel A, varying between .366 and .369. The fixed costs also do not vary between the HPSA and non-HPSA

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<sup>28</sup>We also estimated the parameter  $\sigma$  using only the part of the likelihood function that pertains to the exit and survival flows. In this way the parameter  $\sigma$  is not used to help fit the entry data (through the estimate of  $VE$ ). The estimates of  $\sigma$  were not affected by this change so the fixed cost estimates and the long-run value of the firm can be robustly estimated with or without the data on entry and the potential entry pool. Finally, we also use the nonparametric estimator of  $p^x$  suggested by Pakes, Ostrovsky, and Berry (2007) and find the results are very robust to this alternative.

<sup>29</sup>We also estimated an extension of the model in which the mean of the fixed cost distribution was allowed to vary across markets with differences in the proportion of dentists in the market over age 55. The idea was to see if markets with older dental practices had lower mean fixed costs because their capital equipment had depreciated. The effect was significant and when evaluated at the average value of the age variable produced an estimate of the mean fixed cost identical to the estimate of .308 in Table 4, Panel A. With access to better data on some factors that lead to shifts in the fixed cost distribution across markets it is possible to extend the model to allow for this additional source of market-level heterogeneity.

markets, which is not surprising given the focus of the market subsidy on the entry costs of a new practice.

Given the estimate of  $\sigma = .369$ , we calculate the value of an incumbent continuing in operation,  $VC$ , and the value of entering,  $VE$ , for alternative state vectors  $(n, z, f)$  and these are reported in the top half of Table 5.<sup>30</sup> The estimate of  $VC$ , the discounted sum of expected future net income to the practitioner, varies substantially with the state variables. As we move down each column, increasing the number of firms while holding the exogenous state variables fixed,  $VC$  declines. This reflects two forces: the underlying toughness of short-run competition seen in the slope of the profit function in Figure 1 and the endogenous impact of entry and exit on the long-run firm payoff. As will be discussed below, this latter effect mitigates the decline in long-run profitability arising from the toughness of short-run competition because an increase in the number of firms leads to less entry and more exit in the industry.

Holding  $n$  fixed and allowing the state variables  $(z, f)$  to increase results in substantial increases in  $VC$  as shown in Table 5. This indicates that differences across markets in population, wages, income, and the market fixed effect result in significant differences in long-run firm values, even after accounting for the endogenous effect of entry and exit. For example, a monopoly provider in a market with low-profit characteristics ( $low(z, f)$ ) would have an estimated long-run value of .406 million dollars, while that same monopoly would have a value of 1.168 million dollars in a market with high-profit characteristics. It is clear from comparing the estimates of  $VC$  that differences in exogenous characteristics across markets are more important than differences in the number of firms in determining the long-run value of the firm. The value of entering the market,  $VE$ , is reported in the last three columns of Table 5 and we observe that, at each state vector, the estimates are similar to the estimate of  $VC$  and thus show the same pattern of decline with  $n$  and substantial variation with exogenous market characteristics.<sup>31</sup>

These fixed cost estimates for dentists in Table 4 can also be compared with the estimates

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<sup>30</sup>We construct three combinations of the discrete state variables  $(z_d, f_d)$  which will generate low, medium, and high values of the profit function.

<sup>31</sup>The difference in  $VC$  and  $VE$  arises from the difference between incumbents and entrants in the perceived transition probabilities for the state variables,  $M_e$  and  $\hat{M}_e$ , in equations 9 and 10. These, in turn, differ across the two types of firms because they condition on their own choice. In our application, the estimates of  $\hat{M}_{nc}$  and  $\hat{M}_{ne}$  from equations 13 and 14 are similar, so that the estimates of  $VC$  and  $VE$  are also very similar for each state.

for the chiropractor industry reported in Panel C of Table 4. The estimate of the fixed cost parameter  $\sigma$  is .275 and is not affected by the modeling assumptions we make on the entry cost or entry pool. The estimates of  $VC$  and  $VE$  derived using this value of  $\sigma$  are reported in the bottom half of Table 5. Like the findings for the dentist industry, we see that  $VC$  and  $VE$  both vary substantially with differences in the exogenous state variables  $(z, f)$  and, for a given state,  $VC$  and  $VE$  are very similar in magnitude. These results differ from the dentist findings in two ways. First, the decline in both values as  $n$  increases is not as substantial as the decline for dentists. This partly reflects the earlier finding about the toughness of short-run competition, that an increase in the number of firms has less impact on average profits in this industry, but it will also be affected by how entry and exit respond to the number of firms. Second, the magnitude of  $VC$  and  $VE$  for the chiropractors is substantially less than for dentists. A monopoly dental firm operating in a market with high-profit characteristics would have a firm value of 1.168 million dollars while a monopoly chiropractor in the same type of market would have a firm value of .562 million. This reflects the overall lower level of per-period profit observed for chiropractors.

To more clearly illustrate the variation in  $VC$  across states and the difference in the levels across industries we graph the values of  $VC$  from Table 5 in Figure 2. Each line represents  $VC(n)$  holding the other state variables fixed and thus reflects the endogenous relationship between the number of firms and firm values. The upward shifting of the function reflects the difference due to an increase in the exogenous market characteristics  $(z, f)$ . Finally, to relate the values to the actual data, the size of the circles reflects the number of market/year observations in the data set that have each combination of  $(n, z, f)$ . It is clear from the figure that markets with low  $(z, f)$  values have few firms, while markets with exogenous characteristics that generate higher profits support more firms. However, even in the high profit markets there is wide variation in the number of firms present which implies that some additional source of market heterogeneity, in our case differences in firm fixed costs, sunk entry costs and the number of potential entrants, will be needed to explain the differences in market structure across geographic markets.

Given estimates of the long-run benefits of operating in a geographic market with a given

state, and the fixed cost distribution faced by incumbents, we can estimate the probability of exit and the mean fixed cost faced by surviving firms. Incumbent firms remain in operation if they have a realization of their fixed cost that is less than the value of continuing. Combining equation (2) with the assumption that the fixed cost  $\lambda$  has an exponential distribution, the probability of exit is  $p^x(n, z, f) = \exp(-VC(n, z, f)/\sigma)$ . The first three columns of Table 6 report the estimated probability of exit for each of the states. Reflecting the underlying variation in  $VC$ , the probability of exit rises as the number of firms in the market increases and declines as the exogenous state variables shift toward combinations that result in higher profit states. In the case of dentists, the probability of exit varies from a low of .042 for monopoly markets with high  $(z, f)$  to a high of .866 if a market had 20 firms and low  $(z, f)$  characteristics. In particular there is a large reduction in the exit probability as we move from low to high  $(z, f)$  states. The exit rate in the high  $(z, f)$  states is only one-tenth the magnitude in the low states. The chiropractors have lower values of  $VC$  and a lower value of  $\sigma$  than the distribution for dentists. The former effect will generate higher exit probabilities for chiropractors while the lower  $\sigma$  results in the distribution of fixed costs having more mass on small values which results in lower exit probabilities. The net effect of these two forces, however, always generates predicted exit probabilities that are larger for the chiropractors than for the dentists. The more favorable fixed cost distribution does not compensate for the lower long-run profits and thus there is higher exit in the chiropractor industry.

The mean level of fixed costs incurred by surviving firms depends on both the parameter  $\sigma$  and the truncation point  $VC(n, z, f)$  as shown in equation (4). The first three columns of Table 7 report values of this truncated mean across different states. For example, in the monopoly markets for dentists, if there are low  $(z, f)$  characteristics the monopolist would spend, on average, .166 million dollars on fixed costs and still remain in operation but in high  $(z, f)$  markets spending would average .318 million for the monopoly dentist that remained open. Notice that this occurs, even though the distribution of fixed costs the firms face is identical across all markets, because the amount that operating firms are willing to spend in fixed costs varies with the long-run profitability of the market. Comparing across states, there is more variation in the fixed costs across low, medium, and high  $(z, f)$  markets than across

markets with different numbers of firms. Comparing the two industries the fixed costs that firms would be willing to incur are larger for dentists than chiropractors.

### 6.3 Sunk Costs and the Probability of Entry

The final parameter of interest characterizes the distribution of sunk entry costs faced by potential entrants. In table 4, we report estimates of the entry cost parameter under different assumptions about the shape of the cost distribution and the nature of the potential entrant pool. In Panel A, when we assume that the entry cost distribution is chi-square we get parameter estimates of 2.003 using the internal entry pool definition and 3.286 using the external pool definition. When the entry distribution is exponential the parameter estimates are 2.080 and 8.604 with the internal and external pool definitions, respectively. This dependence on the entry pool definition is not surprising, because as shown in Table 2, the external pool definition generates much larger potential entrant pools and thus lower entry rates in the data. Given the estimates of  $VE$ , which do not depend on the entry cost parameter, the lower entry rates observed with the external pool definition imply a higher level for the entry cost. Focusing on the internal entry pool, the estimated entry cost parameters, 2.003 for the chi-square and 2.080 for the exponential, imply virtually identical entry cost distributions. When the distributions are plotted there is no practical distinction between them. With the external entry pool, the estimates are sensitive to the distributional assumption. When using the exponential distribution there is a higher mean, with less mass on low entry costs and a fatter tail for high entry costs. This sensitivity increases our concerns about the use of the external entry pool definition. When we divide our geographic markets into underserved markets using the HPSA designation a clear distinction arises in the coefficients reported in Panel B of Table 4: in every one of the specifications, the HPSA markets have a lower estimated mean entry cost than the non-HPSA markets. This is consistent with the intent of the policy to encourage practices to locate in these underserved markets. While the level of the mean entry cost depends on the specification of the potential entry pool and entry cost distribution assumption in the same way as seen in Panel A, the lower entry cost for HPSA markets is a robust finding. We will explore the implication of this entry cost difference in counterfactual exercises in the

next section.

For the chiropractor industry, we observe the estimated cost parameter is always smaller than for dentists, regardless of model specification. Comparing the estimates using the internal and external entry pools, the differences are fairly minor: 1.367 for the internal pool and 1.302 for the external pool using the chi-square distributional assumption. This is consistent with the finding in Table 2 that the two definitions do not lead to substantially different measures of the number of the potential entrants. The estimates using the exponential distribution, .920 and .837, is lower than the model using the chi-square assumption but, as was seen for the dentists, plots of the two cost distributions are virtually identical.

Using the mean entry cost estimated using the internal entry pool and exponential cost distribution, we calculate the probabilities of entry using equation (7) and report them in the last three columns of Table 6 for different states. The probability rises as  $(z, f)$  increases and falls as the number of firms increase, reflecting the variation in  $VE$ . The interesting comparison is between the two industries. The distribution of entry costs has a higher mean in the dentist industry but the higher values of  $VE$  lead to a probability of entry that is similar for dentists than chiropractors. For example, the probabilities of entry into a monopoly dentist market are .129, .216, and .343 depending on the level of  $(z, f)$  but are slightly higher, .133, .245, and .371, for monopoly chiropractor markets. This difference between the two industries does increase as the number of firms increases with the chiropractor market having a higher probability of entry in markets with up to 8 firms.

Using the exponential distribution for the entry cost, we also calculate the mean entry cost incurred by the firm's that choose to enter as  $E(\kappa|\kappa < \delta VE) = \alpha - \delta VE(1 - p^e)/p^e$  and these are reported in the last three columns of Table 7.<sup>32</sup> The mean cost varies across states due to variation in both  $VE$  and  $p^e$ . Two patterns are of interest. In high-profit markets, firms will be willing to expend more money to enter. Both the marginal and average entrant into a market will depend on the market characteristics which, in our framework, are the market structure and exogenous profit determinants. Second, the mean realized entry costs are higher in the dentist industry. For example, on average, entrants into the monopoly dentist markets

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<sup>32</sup>The Table 7 estimates use the internal entry pool definition since this is defined in the same way for both industries.

will have sunk costs of .140, .242, or .461 million dollars depending on whether it is a low or high profit market. The entrants into monopoly markets for chiropractors will have average entry costs of .064, .123, and .197 million dollars.<sup>33</sup>

#### 6.4 The Impact of Entry Subsidies: Comparing HPSA and non-HPSA Markets

The goal of this section is to show how long-run profits,  $VC$  and  $VE$ , entry and exit probabilities, and long-run market structure in the dentist industry are driven by the magnitude of the underlying entry cost. Using the parameter estimates in Table 4, we evaluate changes in these endogenous outcomes when the entry cost distribution shifts from the one characterizing the non-HPSA markets to the one characterizing the HPSA markets.<sup>34</sup> We focus on the 59 markets which were classified as HPSA markets during our sample period and use the combinations of states  $(n, z_d, f_d)$  observed in those markets.

Table 8 reports the percentage change in firm values for entering firms and the percentage change in the entry rate and Table 9 reports the same numbers for incumbent firms and the exit rate. The underlying parameter change is a reduction in the mean of the unconditional entry cost distribution from the value estimated for non-HPSA markets ( $\alpha = 2.085$ ) to the value for subsidized HPSA markets ( $\alpha = 1.628$ ), a 22 percent reduction in mean entry costs. We construct the variables for each state and summarize the changes for low, mid, and high  $(z, f)$  markets. In Table 8 we observe that, in the lower cost regime, average entry values  $VE$  for new firms fall between 5.5 and 2.2 percent across states with the largest declines in the markets with few firms. The effect of the increased competition from potential entrants has the largest effect in monopoly markets and declines as the number of incumbent firms in the

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<sup>33</sup>These truncated means are not substantially different if we model entry costs using the chi-square distribution. In this case, the corresponding means for the three profit states for the dentist industry are .121, .213, and .337 and for the chiropractor industry are .051, .098, and .186 million dollars.

<sup>34</sup>Solving the model given alternative values of the entry cost parameter  $\alpha$  requires first that the entrant's and incumbent's optimization problems are solved to give the values of  $VC(n, z_d, f_d)$  and  $VE(n, z_d, f_d)$  at each grid point  $(n, z_d, f_d)$ . We do this using the estimated profit function, equation (11), the empirical transition matrix for  $z$ , equation (15), and the estimated mean fixed costs to simultaneously solve equations (1), (3), and (6). All exercises in this section are done using the parameter estimates in Table 4 for the internal entry pool and the exponential distribution of entry costs. Pakes, Ostrovsky and Berry (2007) provide the formulas for the equilibrium values of a firm's perceptions of the number of entrants and exits for survivors,  $p^c(e, x|n, z, f, \chi^c = 1)$  and entrants,  $p^e(e, x|n, z, f, \chi^e = 1)$ .

market increases. The table also shows that the percentage change in the entry rate is large, between 18 and 22 percent, but does not vary substantially across markets with different states. The corresponding numbers for incumbent firms in Table 9 show a similar decline in the value of the incumbent firms  $VC$ . It shows that the lower entry costs lead to an increase in the exit rate, from 1.5 to 12.6 percent across states, with the largest percentage increase in firm exit coming in the highest profit markets, those with few firms and high  $(z, f)$  values. Markets with few firms are the ones where the oligopoly effect, the decline in average profits as the number of firms increases as seen in Table 3, is most substantial. As expected, a reduction in entry costs acts to increase the competitive pressure from potential entrants which results in both higher firm turnover and lower firm values.

Rather than focusing on the cost of opening a new practice, an alternative policy to increase the number of dental practices in underserved markets would be to subsidize the costs of operating a practice in the market or provide additional payments to the practice based on the number of patients served. Using our model we can assess the impact of policies, such as an operating cost subsidy, that raise firm profits by altering the estimated fixed costs and simulating the change in firm values, entry, exit, and market structure. In order to make these fixed cost simulations comparable to the entry cost simulations just reported, we choose to reduce the mean fixed cost by 8 percent because this generates a very similar change in the distribution of the number of firms. However, the adjustment mechanism is very different in the two cases. The reduction in fixed cost acts to firm raise firm values,  $VC$  and  $VE$ , in most markets, particularly the markets with few firms. Focusing on markets with three or fewer firms,  $VC$  rises, on average, by 1.6, 3.2, and 4.7 percent for the low, mid, and high  $(z, f)$  markets, respectively. This lower fixed cost also reduces the exit rate substantially in these markets with the reduction in the exit rate averaging 9.9, 19.6, and 34.9 percent across the  $(z, f)$  categories. Finally, in contrast to the policy that subsidizes entry costs, the operating cost subsidy has a small effect on the entry rate in these markets, increasing it between 0.5 and 3.7 percent.

Overall, the two subsidies act in very different ways. The entry cost reduction increases the pressure from potential entrants, lowers firm values, and increases turnover, both entry and

exit. The operating cost reduction makes it more profitable to be an incumbent, raises firm values, reduces exit, and leads to a small increase in entry. In both cases we simulate, the 22 percent reduction in entry cost and an eight percent reduction in operating cost, the ultimate impact on market structure is very similar. Table 10 compares the two subsidies in terms of their impact on market structure and the cost per additional firm. The first three rows summarize the distribution of market structure across markets, focusing on the probability of having few firms in a market. Both the entry cost and fixed cost reduction lead to a slight reduction in the number of markets with less than five firms, from .637 in the benchmark case (using costs for the non-HPSA markets) to .609 or .607. Rows 4 through 6 show how this increase is accomplished by summarizing the average number of entrants, exits, and net change in number of firms on average across markets. The entry cost reduction leads to an increase in the average number of entrants per market from 1.026 to 1.257 while the fixed cost reduction leads to 1.059 entrants per market. The number of exiting firms increases when entry costs are reduced, consistent with the change in exit probabilities in Table 8, and decreases when fixed costs are reduced. The net change in the average number of firms per market is a measure of the overall impact of the cost changes on the number of dental firms. In the case of the 22 percent entry cost reduction there is a net increase of .160 firms per market and, when fixed costs are reduced by 8 percent, there is an increase of .163 firms per market. These absolute changes are small, but when compared with the median number of firms in these markets, three, the impact is more substantial.

In addition to affecting entry and exit in very different ways the two cost changes lead to different total expenditures. When entry costs are subsidized, new firms receive the subsidy and the total cost per market is determined by the total number of entrants, but for each 1.257 entrants there is only a net increase of .160 firms. This implies that it takes almost eight entrants to have a net increase of one firm in the market. The increased entry is also accompanied by increased exit so that subsidies are paid to more than just the net number of new firms. This clearly illustrates the difficulty of using entry cost subsidies to increase the number of firms in a market when exit is also endogenous. In contrast, when operating costs are subsidized, all incumbents receive the payment and not just the net number of new firms.

The seventh row of Table 10 reports the average cost of each policy per subsidized market. It equals \$73,000 for the entry cost subsidy and \$70,000 for the fixed cost subsidy. However, because the impact on the actual number of firms per market is small, when expressed as a cost per additional firm in the market the costs of each subsidy are substantial: \$593,000 per additional firm with the entry cost subsidy and \$547,000 with the fixed cost subsidy. The latter figures indicate that it is important to recognize the endogenous response of firms in the market to changes in the subsidy policy.

## 7 Conclusion

Market structure is determined by the entry and exit decisions of individual producers and these are affected by expectations of future profits which, in turn, depend on the nature of competition within the market. In this paper we utilize micro data for two U.S. service industries, dentists and chiropractors, over a 25 year period to study the process of entry and exit and how it determines both market structure and long-run firm values. We estimate a dynamic structural model of firm entry and exit decisions in an oligopolistic industry, based on the model of Pakes, Ostrovsky and Berry (2007), and distinguish the decisions of incumbent firms from potential entrants. We use a panel data set of small geographic markets and data on the average profits of firms and the flows of entering and exiting firms in each market to estimate three underlying structural determinants of entry, exit and long-run profitability. The first is the toughness of short-run price competition, the second is the magnitude of the sunk entry cost faced by potential entrants, and the third is the magnitude of the fixed cost faced by incumbent producers. These three components are treated as the primitives of the model, estimated, and used to measure the distinct impact of incumbents and potential entrants on long-run profitability and market structure.

The results indicate that the toughness of price competition increases with the number of firms. For dental practices the slope of the function  $\pi(n)$  is negative, statistically significant, and particularly large as the number of establishments increases from 1 to 4. In the chiropractor industry the decline is smaller in magnitude but still statistically significant between monopoly and duopoly markets. Estimates of the distributions of entry costs and fixed costs parameters

indicate that they are statistically significant for both industries with the magnitudes being larger in the dental industry. Overall, the estimates indicate that all three primitives of the model are important components of long-run firm values and market structure. As the number of firms in the market increases, the value of continuing in the market and the value of entering the market both decline, the probability of exit rises, and the probability of entry declines. These outcomes also differ substantially across markets due to differences in exogenous cost and demand factors.

For the dental industry we utilize information on a government policy that subsidized entry into underserved markets. In our data 59 of the geographic markets were designated as underserved markets at some point during our sample period. We estimate that the mean entry cost is effectively 22 percent lower in these markets and counterfactual simulations indicate that this cost reduction will lead to approximately .16 more firms per market, on average. The increase in the number of firms reflects both an increase in the number of entrants but also an increase in the amount of exit and a reduction in long run profit. The estimated cost of the subsidy averages approximately \$73,000 per market but almost \$600,000 for each increase of one firm in the long run. In contrast, a subsidy that targets the fixed cost of incumbent firms has virtually the same effect on long-run market structure but a slightly lower cost, \$547,000 per firm. The mechanism in this case, however, is very different because it primarily acts to lower the exit margin while raising the average profit of incumbent firms. The counterfactuals illustrate the difficulty of predicting the impact of policies to increase the number of dental practices in underserved markets when there is both endogenous exit and negative effects of entry on firm profits.

The results reported here also indicate several directions for future research in empirical modeling of entry and exit dynamics. While the estimates of fixed costs and the toughness of short-run competition are not sensitive to modeling assumptions on the pool of potential entrants, the estimates of sunk entry costs are. In this study we treat the pool of potential entrants as exogenous in each market but it would be desirable to better understand what determines variation in the number of potential entrants across markets. Incorporating additional sources of market-level heterogeneity in the distributions of fixed costs or entry costs is a second

area where the basic model can be extended in a straightforward way given the availability of data that would account for across-market shifts in the cost distributions. A third area for research involves incorporating firm-level heterogeneity in profits, fixed costs, and/or entry costs that is correlated over time for individual firms. This would recognize that, for example, a firm that has low idiosyncratic fixed costs in one time period, and is thus unlikely to exit, may have a similar cost structure in future periods. In the model we estimate in this paper, this is less of an issue since our focus is on how entry and exit rates vary across geographic markets with different profit determinants, but it will be important in explaining individual firm patterns of participation or exit.

## References

- [1] Akerberg, D., L. Benkard, S. Berry, and A. Pakes (2007), "Econometric Tools for Analyzing Market Outcomes," in J.J. Heckman and E. Leamer (ed.), *Handbook of Econometrics Vol 6A*, Netherlands: Elsevier Science Publishers.
- [2] Aguirregabiria, V. and P. Mira (2007), "Sequential Estimation of Dynamic Discrete Games," *Econometrica*, Vol. 75, No. 1, pp.1-53.
- [3] American Dental Association (2001), Dental and Medical Education and Licensure in the United States: A Comparison, Council on Dental Education and Licensure.
- [4] American Dental Association (2002), *Survey of Dental Practice: Income From the Private Practice of Dentistry*.
- [5] Aspland, M. and V. Nocke (2006), "Firm Turnover in Imperfectly Competitive Markets," *The Review of Economic Studies*, Vol 73, No. 2, pp.295-327.
- [6] Aw, B.Y., X. Chen, and M.J. Roberts (2001), "Firm-level Evidence on Productivity Differentials and Turnover in Taiwanese Manufacturing," *Journal of Development Economics*, Vol. 66, pp. 51-86.
- [7] Bailey, M.N., C. Hulten, and D. Campbell (1992), "Productivity Dynamics in Manufacturing Plants," *Brookings Papers on Economic Activity: Microeconomics*, Brookings Institution, pp. 187-267.
- [8] Bajari, P., C.L.Benkard, and J. Levin (2007), "Estimating Dynamic Models of Imperfect Competition," *Econometrica*, Vol. 75, No. 5, pp.1331-1370.
- [9] Bartelsman, E., J. Haltiwanger, and S. Scarpetta (2009), "Measuring and Analyzing Cross-Country Differences in Firm Dynamics," in T. Dunne, J.B. Jensen, and M.J. Roberts (eds), *Producer Dynamics: New Evidence from Micro Data*, Chicago: University of Chicago Press.
- [10] Berry, S. (1992), "Estimation of a Model of Entry in the Airline Industry," *Econometrica*, Vol. 60, No. 4, pp. 889-917.

- [11] Berry, S. and P. Reiss (2007), "Empirical Models of Entry and Market Structure," in M. Armstrong and R. Porter (eds.), *Handbook of Industrial Organization, Vol.III*, Netherlands: Elsevier Science Publishers.
- [12] Bresnahan, T. and P. Reiss (1987), "Do Entry Conditions Vary Across Markets?" *Brookings Papers on Economic Activity, Microeconomics Annual, Vol. 1*, pp. 833-882.
- [13] Bresnahan, T. and P. Reiss (1991), "Entry and Competition in Concentrated Markets," *Journal of Political Economy*, Vol. 99, No. 5, pp. 977-1009.
- [14] Campbell, J. and H. Hopenhayn (2005), "Market Size Matters," *Journal of Industrial Economics*, Vol. 53, No. 1. pp. 1-25.
- [15] Caves, R.E. (1998), "Industrial Organization and New Findings on the Turnover and Mobility of Firms," *Journal of Economic Literature*, Vol 36, No. 4, pp.1947-1982.
- [16] Collard-Wexler, A.(2006), "Plant Turnover and Demand Fluctuation in the Ready-Mix Concrete Industry," Center for Economic Studies Working Paper 06-08, U.S. Census Bureau.
- [17] Das, S., M.J. Roberts and J. Tybout (2007), "Market Entry Costs, Producer Heterogeneity, and Export Dynamics," *Econometrica*, Vol. 75, No. 3, pp. 837-873.
- [18] Dixit, A. and R. Pindyck (1994), *Investment Under Uncertainty*, Princeton University Press.T
- [19] Dunne, T., S.D. Klimek, M.J. Roberts, and D.Y.Xu (2009), "The Dynamics of Market Structure and Market Size in Two Health Service Industries," in T. Dunne, J.B. Jensen, and M.J. Roberts (eds), *Producer Dynamics: New Evidence from Micro Data*, Chicago: University of Chicago Press.
- [20] Dunne, T. and M.J. Roberts (1991), "Variation in Producer Turnover Across U.S Manufacturing Industries" in P.A. Geroski and J. Schwalbach (eds.), *Entry and Market Contestability*, Blackwell.

- [21] Dunne, T., M.J. Roberts, and L. Samuelson (1988), "Patterns of Firm Entry and Exit in U.S. Manufacturing Industries," *The Rand Journal of Economics*, Vol. 19, No. 4, pp. 495-515.
- [22] Ericson, R. and A. Pakes (1995), "Markov-Perfect Industry Dynamics: A Framework for Empirical Work," *The Review of Economic Studies*, Vol. 62, No. 1, pp. 53-82.
- [23] Foster, L., J. Haltiwanger, and C.J. Krizan (2001), "Aggregate Productivity Growth: Lessons from Microeconomic Evidence," in E. Dean, M. Harper, and C. Hulten (eds.), *New Contributions to Productivity Analysis*, Chicago: University of Chicago Press.
- [24] Hopenhayn, H. (1992), "Entry, Exit, and Firm Dynamics in Long Run Equilibrium," *Econometrica*, Vol. 60, No. 5, pp.1127-1150.
- [25] Jarmin, R. S., S.D. Klimek, and J. Miranda (2009), "The Role of Retail Chains: National, Regional, and Industry Results," in T. Dunne, J.B. Jensen, and M.J. Roberts (eds), *Producer Dynamics: New Evidence from Micro Data*, Chicago: University of Chicago Press.
- [26] Jarmin, R.S. and J. Miranda (2002), "The Longitudinal Business Database," Center for Economic Studies Working Paper 02-14, U.S. Census Bureau.
- [27] Jovanovic, B (1982), "Selection and the Evolution of Industry," *Econometrica*, Vol. 50, No. 3, pp. 649-670.
- [28] Lambson, V.E. (1991), "Industry Evolution with Sunk costs and Uncertain Market Conditions," *International Journal of Industrial Organization*, Vol. 9, pp. 171-196.
- [29] Mazzeo, M.J. (2002), "Product Choice and Oligopoly Market Structure," *Rand Journal of Economics*, Vol. 33, No. 2, pp.221-242.
- [30] Osterhaus (2006), "Before You Pull the Trigger...Purchasing a Dental Practice," Inscip-tions, Arizona Dental Association.
- [31] Pakes, A., M. Ostrovsky and S. Berry (2007), "Simple Estimators for Parameters of Dis-crete Dynamic Games (with Entry/Exit Examples)," *The Rand Journal of Economics*, Vol. 38, No. 2, pp.373-399.

- [32] Pesendorfer, M. and P. Schmidt-Dengler (2003), "Identification and Estimation of Dynamic Games," NBER Working Paper 9726.
- [33] Ryan, S. (forthcoming), "The Costs of Environmental Regulation in a Concentrated Industry," *Econometrica*.
- [34] Sandefur, R. And I. Coulter (1997), "Licensure and legal Scope of Practice," in *Chiropractic in the United States: Training, Practice, and Research*, Agency for Health Care Policy and Research, 98-N002.
- [35] Seim, K. (2006), "An Empirical Model of Firm Entry with Endogenous Product-Type Choices," *The Rand Journal of Economics*, Vol. 37, No. 3, pp.619-642.
- [36] Sutton, J. (1991), *Sunk Costs and Market Structure*, Cambridge, MA: MIT Press.
- [37] Syverson, C. (2004), "Market Structure and Productivity: A Concrete Example," *The Journal of Political Economy*, vol. 112, No. 6, pp. 1181-1222.
- [38] US Census Bureau (2000), 1997 Business Expenses, Department of Commerce.
- [39] Weaver, R., K. Haden and R. Valachovic (2001), "Annual ADEA Survey of Dental Seniors-2000 Graduating Class," *Journal of Dental Education*, pp. 788-802.

**Table 1: Entry and Exit Statistics**

(means across market-time observations)

<b>Dentist</b>						
Number Estabs	Number Entrants	Number Exits	Entry Proportion	Exit Rate	% of obs with E,X > 0	
<i>n</i> =1	.489	.285	.489	.285	.102	
<i>n</i> =2	.528	.345	.264	.172	.206	
<i>n</i> =3	.700	.522	.233	.174	.289	
<i>n</i> =4	.799	.675	.199	.169	.318	
<i>n</i> =5	.939	.893	.188	.179	.412	
<i>n</i> =6	1.18	1.09	.196	.182	.472	
<i>n</i> =7	1.34	1.43	.191	.203	.622	
<i>n</i> =8	1.57	1.58	.195	.198	.661	
<i>n</i> =9	1.74	1.76	.192	.196	.750	
<i>n</i> =10,11	1.99	2.17	.191	.208	.785	
<i>n</i> =12,13,14	2.62	2.51	.202	.194	.853	
<i>n</i> =15,16,17	2.85	3.03	.180	.191	.915	
<i>n</i> =18,19,20	3.31	4.36	.176	.231	.984	
<b>Chiropractor</b>						
<i>n</i> =1	.847	.229	.847	.229	.197	
<i>n</i> =2	.932	.436	.466	.218	.279	
<i>n</i> =3	1.02	.744	.341	.248	.408	
<i>n</i> =4	1.40	1.03	.350	.256	.532	
<i>n</i> =5	1.59	1.37	.319	.273	.669	
<i>n</i> =6	1.91	1.83	.319	.304	.757	
<i>n</i> =7	2.05	2.26	.292	.322	.744	
<i>n</i> =8	3.25	2.92	.406	.365	.917	

**Table 2: Number of Potential Entrants**

(mean across market-time observations)

Number of Estabs	<b>Dentists</b>		<b>Chiropractors</b>	
	Number of Potential Entrants		Number of Potential Entrants	
	internal entry pool	external entry pool	internal entry pool	external entry pool
<i>n</i> =1	2.31	23.55	3.42	1.95
<i>n</i> =2	2.74	25.22	3.78	2.88
<i>n</i> =3	3.48	23.41	4.25	4.21
<i>n</i> =4	4.04	23.05	5.13	5.37
<i>n</i> =5	4.75	23.79	5.61	6.83
<i>n</i> =6	6.03	25.45	6.19	7.74
<i>n</i> =7	6.58	27.83	6.16	9.37
<i>n</i> =8	7.81	29.09	8.75	10.67
<i>n</i> =9	8.53	28.26		
<i>n</i> =10,11	9.66	27.13		
<i>n</i> =12,13,14	11.74	25.89		
<i>n</i> =15,16,17	13.83	27.15		
<i>n</i> =18,19,20	15.95	28.21		

**Table 3: Profit Function Parameter Estimates**

(standard deviation in parentheses)

Dentist			Chiropractor		
Variable	No Market Fixed Effect	Market Fixed Effect	Variable	No Market Fixed Effect	Market Fixed Effect
<i>Intercept</i>	-8.256* (4.113)	-1.132 (4.027)	<i>Intercept</i>	-10.658 (6.057)	-27.651** (7.596)
<i>I(n = 1)</i>	.0354 (.0339)	.0582 (.0310)	<i>I(n = 1)</i>	.0157 (.0175)	.0554 (.0371)
<i>I(n = 2)</i>	.0244 (.0173)	.0395 (.0227)	<i>I(n = 2)</i>	.0178 (.0166)	.0347 (.0363)
<i>I(n = 3)</i>	.0125 (.0123)	.0221 (.0168)	<i>I(n = 3)</i>	.0066 (.0158)	.0275 (.0358)
<i>I(n = 4)</i>	.0106 (.0102)	.0144 (.0125)	<i>I(n = 4)</i>	.0023 (.0153)	.0146 (.0352)
<i>I(n = 5)</i>	.0203* (.008)	.0164 (.009)	<i>I(n = 5)</i>	-.0002 (.0164)	.0303 (.0358)
<i>n</i>	-.003 (.004)	-.0223** (.006)	<i>I(n = 6)</i>	.0035 (.0174)	.0053 (.0361)
<i>n</i> <sup>2</sup>	.0001 (.0001)	.0005* (.0002)	<i>I(n = 7)</i>	-.0247 (.0163)	-.0196 (.0368)
<i>pop</i>	.0003 (.0169)	-.0186 (.0244)	<i>pop</i>	-.0122 (.0201)	-.0449 (.0278)
<i>pop</i> <sup>2</sup>	-.0001 (.0000)	-.0002 (.0001)	<i>pop</i> <sup>2</sup>	-.0001 (.0000)	-.0001 (.0001)
<i>inc</i>	1.851* (.8983)	-.0041* (.8858)	<i>inc</i>	2.150 (1.313)	5.823** (1.648)
<i>inc</i> <sup>2</sup>	-.1029* (.0492)	.0148 (.0002)	<i>inc</i> <sup>2</sup>	-.1072 (.0712)	-.3056** (.0895)
<i>w</i>	-.2149** (.0623)	-.1255** (.0515)	<i>w</i>	.2417** (.0817)	.1589** (.0654)
<i>w</i> <sup>2</sup>	.0013* (.0002)	-.0009** (.0002)	<i>w</i> <sup>2</sup>	-.0005** (.0002)	-.0003* (.0001)
<i>pop * w</i>	.0002 (.0001)	-.0001 (.0002)	<i>pop * w</i>	.0001 (.0001)	.0001 (.0001)
<i>pop * inc</i>	.0004 (.0018)	.0041 (.0027)	<i>pop * inc</i>	.0016 (.0022)	.0053 (.0357)
<i>inc * w</i>	.0272 (.0069)	.0159** (.0056)	<i>inc * w</i>	-.0237** (.0089)	-.0156* (.0071)

\*\* significant at the .01 level, \* significant at .05 level

**Table 4: Fixed Cost and Entry Cost Parameter Estimates**

(standard errors in parentheses)

Chi-square Entry Cost			Exponential Entry Cost			
<b>Panel A. Dentist (All Markets)</b>						
Entry pool	$\sigma$	$\alpha$		$\sigma$	$\alpha$	
internal	0.370 (0.005)	2.003 (0.016)		0.369 (0.005)	2.080 (0.041)	
external	0.373 (0.005)	3.286 (0.056)		0.369 (0.005)	8.604 (0.423)	
<b>Panel B. Dentist (HPSA vs Non-HPSA Markets)</b>						
Entry pool	$\sigma$	$\alpha$ (HPSA)	$\alpha$ (non-HPSA)	$\sigma$	$\alpha$ (HPSA)	$\alpha$ (non-HPSA)
internal	0.366 (0.009)	1.804 (0.072)	2.016 (0.048)	0.366 (0.008)	1.628 (0.101)	2.085 (0.103)
external	0.369 (0.006)	3.054 (0.182)	3.470 (0.109)	0.366 (0.008)	7.117 (1.683)	10.522 (0.979)
<b>Panel C. Chiropractor</b>						
Entry pool	$\sigma$	$\alpha$		$\sigma$	$\alpha$	
internal	0.275 (0.005)	1.367 (0.015)		0.275 (0.005)	0.920 (0.017)	
external	0.274 (0.005)	1.302 (0.022)		0.275 (0.005)	0.837 (0.036)	

**Table 5: Predicted Value of Dynamic Benefits VC, VE**

(evaluated at different values of the state variables)

(millions of 1983 dollars)

	<i>VC</i> for Incumbents - Dentist			<i>VE</i> for Potential Entrants - Dentist		
	low( $z, f$ )	mid( $z, f$ )	high( $z, f$ )	low( $z, f$ )	mid( $z, f$ )	high( $z, f$ )
$n=1$	.406	.693	1.168	.370	.653	1.129
$n=2$	.359	.646	1.121	.329	.612	1.087
$n=3$	.310	.598	1.073	.286	.569	1.045
$n=4$	.279	.566	1.042	.258	.541	1.016
$n=5$	.245	.532	1.008	.227	.511	.986
$n=6$	.222	.509	.985	.206	.490	.965
$n=7$	.202	.489	.964	.188	.471	.946
$n=8$	.183	.470	.946	.171	.454	.929
$n=10$	.148	.435	.911	.138	.421	.896
$n=12$	.117	.404	.880	.110	.393	.868
$n=16$	.079	.366	.841	.074	.357	.832
$n=20$	.053	.340	.816	.051	.334	.809
	<i>VC</i> for Incumbents - Chiro			<i>VE</i> for Potential Entrants - Chiro		
	low( $z, f$ )	mid( $z, f$ )	high( $z, f$ )	low( $z, f$ )	mid( $z, f$ )	high( $z, f$ )
$n=1$	.178	.344	.562	.170	.335	.553
$n=2$	.166	.332	.551	.161	.326	.544
$n=3$	.155	.321	.540	.151	.316	.534
$n=4$	.148	.314	.532	.144	.308	.527
$n=5$	.138	.304	.522	.134	.299	.517
$n=6$	.132	.298	.516	.129	.294	.512
$n=7$	.127	.294	.512	.126	.291	.509
$n=8$	.123	.289	.508	.123	.287	.506

**Table 6: Predicted Probabilities of Exit and Entry**

(evaluated at different values of the state variables)

	Probability of Exit - Dentist			Probability of Entry - Dentist		
	low( $z, f$ )	mid( $z, f$ )	high( $z, f$ )	low( $z, f$ )	mid( $z, f$ )	high( $z, f$ )
$n=1$	.333	.153	.042	.129	.216	.343
$n=2$	.378	.174	.048	.115	.204	.332
$n=3$	.431	.198	.055	.101	.191	.322
$n=4$	.470	.216	.059	.091	.182	.315
$n=5$	.515	.236	.065	.081	.173	.307
$n=6$	.548	.252	.069	.074	.166	.301
$n=7$	.579	.266	.073	.067	.161	.296
$n=8$	.608	.279	.077	.061	.155	.292
$n=10$	.669	.307	.085	.050	.145	.283
$n=12$	.728	.334	.092	.039	.136	.276
$n=16$	.808	.371	.102	.027	.124	.266
$n=20$	.866	.398	.110	.019	.117	.260'
	Probability of Exit - Chiro			Probability of Entry - Chiro		
	low( $z, f$ )	mid( $z, f$ )	high( $z, f$ )	low( $z, f$ )	mid( $z, f$ )	high( $z, f$ )
$n=1$	.524	.286	.129	.133	.245	.371
$n=2$	.547	.299	.135	.127	.239	.367
$n=3$	.569	.311	.141	.119	.233	.362
$n=4$	.585	.319	.144	.114	.228	.358
$n=5$	.606	.331	.150	.107	.222	.352
$n=6$	.620	.339	.153	.103	.219	.350
$n=7$	.629	.344	.155	.101	.217	.348
$n=8$	.639	.349	.158	.098	.215	.346

**Table 7: Estimated Fixed and Sunk Costs**  
 (evaluated at different values of the state variables)  
 (millions of 1983 dollars)

	Mean Fixed Cost for Survivors - Dentist			Mean Sunk Cost for Entrants - Dentist		
	low( $z, f$ )	mid( $z, f$ )	high( $z, f$ )	low( $z, f$ )	mid( $z, f$ )	high( $z, f$ )
$n=1$	.166	.243	.318	.140	.242	.406
$n=2$	.151	.233	.313	.125	.228	.392
$n=3$	.134	.221	.307	.109	.212	.378
$n=4$	.122	.213	.303	.098	.202	.368
$n=5$	.109	.204	.299	.087	.191	.358
$n=6$	.100	.198	.296	.079	.184	.351
$n=7$	.092	.192	.293	.072	.177	.344
$n=8$	.084	.187	.290	.065	.171	.338
$n=10$	.069	.176	.285	.053	.159	.327
$n=12$	.056	.166	.280	.042	.148	.317
$n=16$	.038	.153	.273	.028	.135	.305
$n=20$	.026	.144	.268	.020	.126	.297
	Mean Fixed Cost for Survivors - Chiro			Mean Sunk Cost for Entrants - Chiro		
$n=1$	.079	.137	.191	.064	.123	.197
$n=2$	.075	.133	.189	.061	.120	.194
$n=3$	.070	.130	.187	.057	.117	.191
$n=4$	.067	.128	.185	.054	.114	.189
$n=5$	.063	.124	.183	.051	.111	.185
$n=6$	.061	.123	.182	.049	.109	.183
$n=7$	.059	.121	.181	.048	.108	.183
$n=8$	.057	.120	.180	.047	.107	.182

**Table 8: Reduction in Entry Cost: Impact on Entrants**

(percentage change in the variable)

Number of Firms	$VE(n, z, f)$			$p^e(n, z, f)$		
	Low ( $z, f$ )	Mid ( $z, f$ )	High ( $z, f$ )	Low ( $z, f$ )	Mid ( $z, f$ )	High ( $z, f$ )
$n = 1$	-5.6	-4.2	-3.1	19.2	19.4	18.3
$n = 2$	-5.1	-3.9	-2.9	20.0	20.0	18.8
$n = 3$	-5.2	-3.8	-2.8	20.0	20.3	19.0
$n = 4$	-4.8	-3.6	-2.7	20.6	20.7	19.3
$n = 5$	-4.8	-3.5	-2.6	20.8	20.9	19.5
$n = 7$	-4.1	-3.1	-2.4	21.8	21.6	19.8
$n = 9$	-3.4	-2.8	-2.2	22.8	22.1	20.3

**Table 9: Reduction in Entry Cost: Impact on Incumbent Firms**

(percentage change in the variable)

Number of Firms	$VC(n, z, f)$			$p^x(n, z, f)$		
	Low ( $z, f$ )	Mid ( $z, f$ )	High ( $z, f$ )	Low ( $z, f$ )	Mid ( $z, f$ )	High ( $z, f$ )
$n = 1$	-6.4	-4.9	-3.7	7.5	9.9	12.6
$n = 2$	-5.9	-4.5	-3.4	6.0	8.3	11.1
$n = 3$	-5.8	-4.4	-3.2	5.1	7.5	10.1
$n = 4$	-5.4	-4.1	-3.0	4.2	6.6	9.2
$n = 5$	-5.3	-4.0	-3.0	3.7	6.0	8.5
$n = 7$	-4.5	-3.5	-2.7	2.5	4.8	7.4
$n = 9$	-3.7	-3.1	-2.4	1.7	4.0	6.5

**Table 10: Cost-Benefit Comparison of Entry Cost and Fixed Cost Subsidy**

	Benchmark	Entry Cost Reduction	Fixed Cost Reduction
Impact on Market Structure			
Pr( $n = 1$ )	.079	.069	.070
Pr( $n \leq 3$ )	.379	.350	.351
Pr( $n \leq 5$ )	.637	.609	.607
Av. Number of Entrants/Market	1.026	1.257	1.059
Av. Number of Exits/Market	0.989	1.097	0.896
Net Change in Firms/Market	0.037	0.160	0.163
Cost/Market (million \$)		.073	.070
Cost/Additional Firm (millions \$)		.593	.547

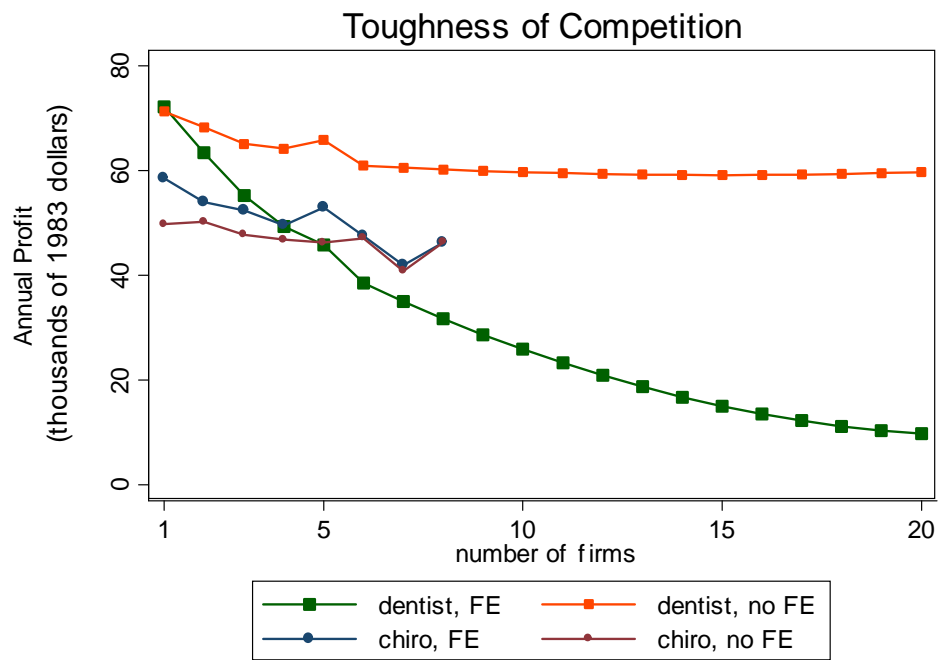


Figure 1

# Value of Continuation- $VC(n, z, f)$

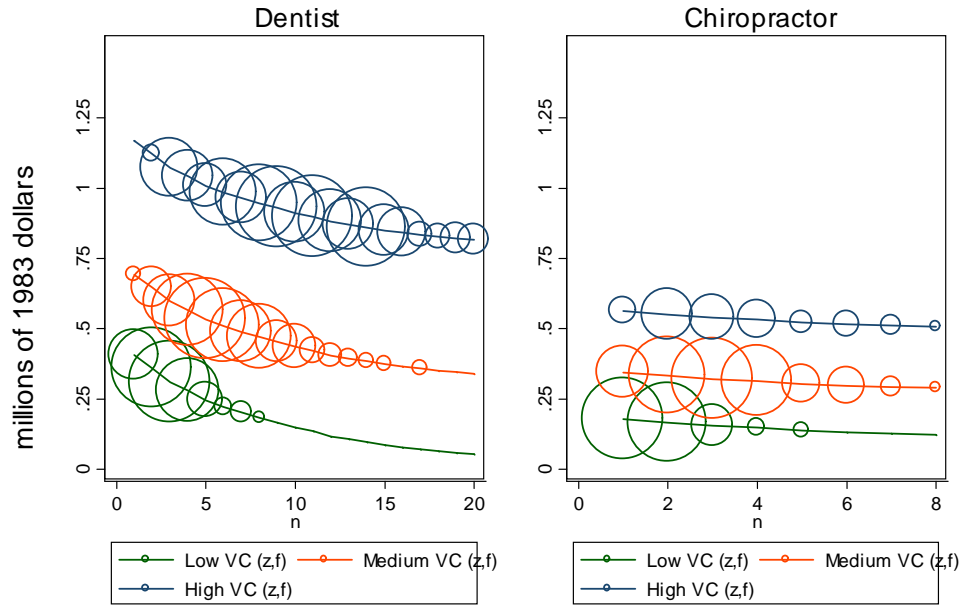


Figure 2