

Regular Adjustment: Theory and Evidence

Jerzy Konieczny and Fabio Rumler

June 2007

This paper is part of the Kiel Working Paper Collection No. 2

“The Phillips Curve and the Natural Rate of Unemployment”

June 2007

http://www.ifw-kiel.de/pub/kap/kapcoll/kapcoll_02.htm

Kiel Institute for World Economics
Duesternbrooker Weg 120
24105 Kiel (Germany)

Kiel Working Paper No. 1352

Regular Adjustment: Theory and Evidence

by

Jerzy Konieczny and Fabio Rumler

June 2007

The responsibility for the contents of the working papers rests with the authors, not the Institute. Since working papers are of a preliminary nature, it may be useful to contact the authors of a particular working paper about results or caveats before referring to, or quoting, a paper. Any comments on working papers should be sent directly to the authors.

Regular Adjustment: Theory and Evidence

Jerzy (Jurek) D. Konieczny
Department of Economics
Wilfrid Laurier University
Waterloo, Ont., Canada, N2L 3C5
jurek@wlu.ca

Fabio Rumler
Oesterreichische Nationalbank
Otto Wagner Platz 3,
A-1090 Vienna, Austria
Fabio.Rumler@oenb.at

July 14, 2006

Abstract

We ask why, in many circumstances and many environments, decision-makers choose to act on a *time-regular* basis (e.g. adjust every six weeks) or on a *state-regular* basis (e.g. set prices ending in a 9), even though such an approach appears suboptimal. The paper attributes regular behaviour to adjustment cost heterogeneity. We show that, given the cost heterogeneity, the likelihood of adopting regular policies depends on the shape of the benefit function: the flatter it is, the more likely, *ceteris paribus*, is regular adjustment. We provide sufficient conditions under which, when policymakers differ with respect to the shape of the benefit function (as in Konieczny and Skrzypacz, 2006), the frequency of adjustments across markets is negatively correlated with the incidence of regular adjustments. On the other hand, if policymakers differences are due to the level of adjustment costs (as in Dotsey, King and Wolman, 1999), then the correlation is positive.

To test the model we apply it to optimal pricing policies. We use a large Austrian data set, which consists of the direct price information collected by the statistical office and covers 80% of the CPI over eight years. We run cross-sectional tests, regressing the proportion of attractive prices and, separately, the excess proportion of price changes at the beginning of a year and at the beginning of a quarter, on various conditional frequencies of adjustment, inflation and its variability, dummies for good types, and other relevant variables. We find that the lower is, in a given market, the conditional frequency of price changes, the higher is the incidence of time- and state-regular adjustment.

JEL codes: E31, L11, E52, D01

Keywords: Optimal pricing, attractive prices, menu costs

We would like to thank the participants in the Eurosystem Inflation Persistence Network (IPN), seminar participants at the Oesterreichische Nationalbank, Wilfrid Laurier University and ECARES, Université Libre, Bruxelles and an anonymous referee for helpful comments and suggestions. We are responsible for any remaining errors.

I. Introduction

“The [FOMC] committee agreed unanimously to lift its benchmark federal funds rate [...] its 16th consecutive quarter-percentage-point increase since June 2004” *The Washington Post*, May 11, 2006.

In many circumstances and in many environments, decision-makers choose to act on a regular basis and, in particular, on a calendar-regular basis (e.g. once a week, on the first day of each quarter, etc.) even though such an approach appears suboptimal. Similarly, some decision-makers appear to prefer some values of the variables under their control (e.g. prices ending with a 9, interest rates which are multiples of 0.25% etc.). The focus of this paper is to analyze a simple explanation of such behaviour.

A common feature of the environments in question is their dynamic structure. The policymaker(s) maximizes a stream of benefits, which depends on the values of some state variables. Over time these values change, or *deteriorate*.¹ The policymaker can reset the state variables but doing so involves a cost. Therefore adjustment is infrequent.

The motivation, and the focus of the paper, is the analysis of nominal price adjustment at the firm level. In this application, a firm posts the nominal price for the product(s) it sells. Due to general inflation the real price falls over time. The real price can be reset by choosing and posting a new value of the nominal price. Similar problems arise in many other environments. Therefore we begin by describing issues related to regular adjustment using examples from various potential applications.

1. *Wage adjustment*. Under general inflation, the purchasing power of contractually-set wages declines over time. It can be increased in a new contract.
2. *Machinery refurbishing*. The capital stock deteriorates over time due to physical use or obsolescence. It is improved by refurbishing or replacing the machinery.
3. *Inventory reordering*. A firm holds an inventory of the product(s) it sells. The level of the inventory falls over time. It is replenished by a new delivery.

¹ Alternatively, the current values of the state variables are constant while the optimal values drift over time. These problems are similar and so we will focus mostly on environments with constant optimal values.

4. *Monetary policy.* The Central Bank sets the interest rate appropriate for the current conditions. Over time the match between the current and the optimal value deteriorates. The interest rate can be readjusted through a decision of the Bank's policy-making body.

5. *Fiscal policy.* The fiscal authority sets spending and taxation priorities in the budget. Over time the desired fiscal structure changes. It is reset in a new budget.

6. *Information.* Newspapers and magazines allow the public to update their information. New events lead to its deterioration. A new issue brings the information up to date.

7. *Monitoring patients.* A patient's visit allows the physician to undertake a proper course of action. Over time the health of the patient or the effectiveness of the treatment may decline. A repeat visit allows the doctor to review and adjust the treatment.

These problems are fairly common. As discussed below, they often lead to *state-contingent* adjustment policies. The decision maker monitors the state variable and applies the control whenever it has deteriorated to the threshold point. Hence the timing of adjustment does not depend solely on time and, in general, adjustments are not regular.

In practice, however, we observe many cases where controls are applied at regular moments of time. US grocery stores adjust prices on Wednesdays (Levy et al., 1997); drugstores adjust prices on Fridays (Dutta et al., 1999). Seasonal sales are held every January and July. Many firms get regular deliveries. Machinery is often refurbished on a regular basis. Labour contracts are signed for a fixed number of years. Magazines and newspapers appear with fixed frequency. Medical associations provide guidelines on the frequency of checkups and so on.

In many cases some decision-makers follow regular policies while others do not. While some firms change prices at predetermined dates, others follow *state-contingent* optimal pricing policies (Cecchetti, 1986). Observed hazard functions of price changes in the euro area countries suggest a coexistence of *state-contingent* and *time-regular* price setting² (Álvarez et al., 2005). Car firms change prices in the fall but offer incentives on a *state-contingent* basis (depending on inventory levels). Machinery is often refurbished when predetermined technical requirements are met. Some firms follow just-in-time delivery schedules, etc.

² What we call *time-regular* policy is usually called a *time-contingent* policy. For clarity we avoid the latter term; this allows us to distinguish between *state-regular* and *state-contingent* policies.

Even when the policy is formally regular, it sometimes contains specific provisions for deviating from the schedule if needed. The interest rate may be changed between the regular meetings of the policy makers; the government may introduce a mini-budget and so on.

Furthermore, policymakers sometimes switch between regular and irregular policies. Several years ago the Bank of Canada officially moved from weekly to less frequent meetings. As implied by the above quote, in June 2004 the FED implicitly switched to regular adjustments every six weeks by 0.25%.³ Car producers have switched to just-in-time delivery policies. Most airlines nowadays use sophisticated pricing schedules, etc.

Finally, some policymakers follow different policies for different activities. Paper versions of newspapers are published regularly, but electronic versions are not.⁴ Some supplies may be obtained regularly while others are procured on just-in-time basis. Doctors set regular, routine visits for some patients but not for others, etc.

Understanding of regular policies is important since such policies reduce flexibility by limiting the ability of the policymaker to react to past, current and future events. It is important to note that the distinction between expected and unexpected events is not crucial here. Once the system is set up to adjust on a regular basis, the policymaker may not be able to alter the course of action for a range of both expected and unexpected changes. For example, a central bank which precommits itself to changing the interest rate on a regular basis may be unwilling to break the pattern in the face of either expected or unexpected events.

Explanations of these phenomena depend on the environment. Regular scheduling obviously reduces the cost of maintenance or of inventory delivery. Regular price adjustment may have strategic benefits (avoiding price war) or reputational benefits (easier acceptance by customers).⁵ Regular scheduling of monetary policy decisions helps “reducing uncertainty in the financial markets...” and “...fixed dates will allow market participants to plan and operate more efficiently.”⁶ Regular publishing of magazines is convenient for readers. Guidelines on the frequency of checkups simplify physicians’ decisions, etc.

³ In the previous 16 meetings (June 2002-May 2004) the interest rate was changed once by 0.5%, once by 0.25% and was unchanged 14 times; during the June 2000-May 2002 period it was changed eight times by 0.5%, three times by 0.25% and was left unchanged five times.

⁴ We are grateful to Magdalena Konieczna for suggesting this example.

⁵ See Rotemberg (2005) and (2006) for reputation – based adjustment models.

⁶ Bank of Canada (2000).

Given the variety of environments and motives for adopting regular policies, in this paper we ask whether they can be accounted for with a simple, uniform framework. The model we use assumes that adjustment of the state variable is costly, but the adjustment costs are heterogeneous: they vary over time or over the values of the state variables. When the lower values of the costs occur regularly, for some policymakers regular adjustment dominates the state-contingent policy that would have been optimal if costs were homogeneous.

In order to avoid misunderstanding we want to emphasize two points. First, the adoption of this simple assumption does not mean we argue that adjustment costs are, in fact, heterogeneous in a regular manner. Second, the proposed explanation is by no means trivial.

With regard to the first issue, we treat the assumption of regularly heterogeneous adjustment costs as a simple approach to a complex problem. While applicable in some environments, this assumption is problematic in others. For example, the average unit delivery cost is likely to be lower when the firm prearranges delivery of x truckloads every y weeks rather than order inventory as needed. On the other hand, it is not clear what reduction in costs is obtained by making interest rate decisions four times a year (as the Swiss National Bank does), or by 0.25% (as the FED has been doing). Furthermore, we adopt the simplest assumption possible: we assume that the cost of adjustment is lump-sum and takes on only two values: high and low. We do not claim that this extreme simplification is realistic, but rather ask whether, with this assumption, our model can generate observed behaviour. The answer is a clear yes.

Using the assumption of heterogeneous costs may, at first thought, make our model appear trivial. As our analysis shows, however, that is not the case. We show that the results hold for an arbitrarily small difference between the high and low values of the costs. Furthermore, heterogeneous costs are neither sufficient nor necessary to explain the incidence of regular behaviour. Additional assumptions are needed to obtain testable predictions.

There are two aspects of regular nominal price adjustment we are interested in: time-regularity and state-regularity. A disproportionate proportion of price changes take place at the beginning of periods, rather than within periods. Several studies in the Inflation Persistence Network (IPN) report a high proportion of prices are held constant for a year (Álvarez et al., 2005 for Spain, Aucremanne and Dhyne, 2005 for

Belgium, Baudry et al., 2004 for France, Baumgartner et al., 2005 for Austria, Dias et al., 2005 for Portugal, Veronese et al., 2005 for Italy, Lünnemann and Mathä, 2005 for Luxembourg and Hoffmann and Kurz-Kim, 2005 for Germany). Konieczny and Skrzypacz (2005) report that, in price data collected three times a month, over a half of all changes take place in the first 10 days of a month. Similarly, several IPN studies, as well as Bergen et al. (2003) and Konieczny and Skrzypacz (2006) find that a large proportion of prices charged are *attractive prices*.⁷

Consistent with our approach, several studies on price adjustment have recently addressed the idea of heterogeneity in adjustment costs. Levy et al. (2005) explain heterogeneity in price rigidity across holiday and non-holiday periods by variations in the cost of price adjustment. The papers by Owen and Trzepacz (2000) and by Levy et al. (2002) also contain discussions along these lines. Dotsey, King and Wolman (1999) as well as Wolman (2000) consider cross-product variation in the cost of price adjustment.

We start the paper by showing an existence result: when the costs of adjustment are lower at regular moments of time, and even when the difference is arbitrarily small, an optimizing policymaker will (except in unlikely circumstances) take advantage of the lower costs. We then show that, given the cost heterogeneity, the likelihood of adopting regular policies depends on the shape of the benefit function: the flatter it is, the more likely, *ceteris paribus*, is regular adjustment. In general, however, there is no clear relationship between the degree of cost heterogeneity and the incidence of regular adjustment. In order to obtain empirical predictions we add heterogeneity across policymakers. We consider two sources or differences across policymakers: the shape of the benefit function (as in Konieczny and Skrzypacz, 2006 and the size of the adjustment costs as in Dotsey, King and Wolman, 1999). We provide sufficient conditions under which, with the differences across policymakers being due to the differences in the shape of the benefit function, the frequency of adjustments across markets is negatively correlated with the incidence of regular adjustments. On the other hand, if the differences across policymakers are due to the level of adjustment costs, the correlation is positive.

⁷ Attractive prices – which sometimes are also called threshold prices or pricing points – include psychological prices (prices ending in 9), fractional prices (prices which are convenient to pay, such as 1.50) and round prices (defined as whole number amounts, such as 10.00).

We then apply the model to nominal price adjustment. The distinction between the time contingent, regular nominal price adjustment policies (as in Fischer, 1977 and in Taylor, 1980), and state-contingent policies (as in Sheshinski and Weiss, 1977), is crucial, given their different implications for effectiveness of monetary policy (Caplin and Spulber, 1987, Caplin and Leahy, 1992).

To test the model we use a very large Austrian data set, which consists of the direct price information collected by the statistical office and covers about 80% of the CPI over eight years. We run cross-sectional tests, regressing the proportion of attractive prices and, separately, the excess proportion of price changes at the beginning of a year and at the beginning of a quarter on various conditional frequencies of adjustment, inflation and its variability, dummies for good types, and other relevant variables. We find that the lower is, in a given market, the conditional frequency of price changes, the higher is the incidence of time- and state- regular adjustment. This is consistent with markets being heterogeneous with respect to the shape of the profit function, but not consistent with markets differing with respect to the value of the menu costs.

The paper is organized as follows. The model is introduced, and the empirical predictions are derived in the next section. In section 3 we discuss the empirical evidence. Conclusions are in the last section.

II. The Model.

We consider a class of optimization problems where the value of instantaneous benefits depends on state variables that change over time. More formally, the instantaneous value of the benefits is $\tilde{B}[\bar{x}(t), \bar{y}(t), \bar{a}]$, where $\bar{x}(t)$ is a vector of state variables, $\bar{y}(t)$ is a vector of exogenous variables and \bar{a} is a vector of parameters. This formulation implies that the benefit function depends on time only indirectly.

We assume that $\tilde{B}[\bar{x}(t), \bar{y}(t), \bar{a}]$ is twice continuously differentiable and has a unique global maximum:

A1. For every t , $\bar{y}(t), \bar{a}$ there exists $\bar{x}^*(\bar{y}(t), \bar{a})$ such that, for every $\bar{x}(t) \neq \bar{x}^*$:

$$\tilde{B}[\bar{x}(t), \bar{y}(t), \bar{a}] < \tilde{B}[\bar{x}^*, \bar{y}(t), \bar{a}]$$

Assumption A1 implies that, as long as \bar{y} and \bar{a} do not change, the optimal instantaneous values of the state variables are constant.

The policymaker would like to maintain the state variables continuously at the level \bar{x}^* or, if that is not possible, to keep them close to \bar{x}^* . Changes in $\bar{x}(t)$ over time will be called the *deterioration* of the state variables. The policy maker can adjust $\bar{x}(t)$ at any time to any desired level (perhaps within some bounds), but doing so involves a discrete cost.⁸

The cost of adjusting the state variable, suggested by the examples above, includes the time, or the opportunity cost of the time needed to set up the decision-making process (e.g. organizing an election and counting votes, the doctor's and the patient's time etc.), the time needed to make and implement the decision (e.g. the time needed to set up and implement a new budget, union/employer bargaining time etc.), physical resources (e.g. new machinery, printing a new price list etc.) and non-time opportunity costs (e.g. foregone output whenever production is affected by the refurbishing process etc.).

To simplify the analysis, and in line with earlier literature (Scarf, 1959, Sheshinski and Weiss, 1977), we assume that the cost is lump-sum: independent of the size or of the frequency of adjustment. This is a reasonable assumption in some cases (monetary policy decisions, printing a new price list etc.).⁹

In general, the optimal solution to the optimization problems described above is *state-contingent*. The policymaker observes the values of the state variables and, when they reach certain thresholds, incurs the discrete cost and adjusts them to new, optimally chosen levels. State-contingent policies imply, generally, adjustment at intervals of differing length. Thresholds, as well as the new values of state variables are computed optimally and can take on any values (from an admissible range).

As discussed in the introduction, in many environments, however, we observe behaviour inconsistent with state-contingent policies: adjustment often takes place at regular intervals and some values of the state variables are chosen more often than others. We focus, therefore, on adjustment policies which we call regular policies. We distinguish between *time-regular* policies, which involve adjustment on a regular

⁸ In an equivalent problem, the optimal values change over time and the goal of the policymaker is to maintain the state variable as close as possible to the drifting optimal value, given the adjustment costs.

⁹ Adjustment costs often include, in addition, a component which depends on the size of adjustment (refurbishing machinery, delivering a mini-budget etc.). We do not consider such cases here.

basis (e.g. a firm orders new inventory every 48 days, monetary policy decision making body meets every six weeks, machinery is refurbished once every sixteen months etc.) and *state-regular* policies, in which newly chosen values of the state variables belong to a small subset of all possible values (e.g. inventory is ordered by a truckload, a firm selects new prices ending in a nine: 0.69, 0.79 etc.). An important subset of time-regular policies are *calendar time-regular* policies, which involve adjustment at calendar-related intervals (e.g. a new price list is issued once a year etc.) or where the time of applying the control is related to the calendar (e.g. sales are held at the beginning of each January and each July)

To make the analysis tractable we make several simplifying assumptions:

- A2. Over the relevant range, and for any values of $\bar{y}(t), \bar{a}$, the effect of the vector $\bar{x}(t)$ on the benefit function $\tilde{B}[\bar{x}(t), \bar{y}(t), \bar{a}]$ can be completely summarized by a single state variable $x(t)$.¹⁰ i.e. there exists $B[.]$ such that $B[x(t), \bar{y}(t), \bar{a}] \equiv \tilde{B}[\bar{x}(t), \bar{y}(t), \bar{a}]$ and, for every $t, \bar{y}(t), \bar{a}$ there exists $x^*(\bar{y}(t), \bar{a})$ such that, for every $x(t) \neq x^*$: $B'[x(t), \bar{y}(t), \bar{a}] \cdot [x^* - x(t)] < 0$. where $B'[.]$ denotes the derivative of the benefit function with respect to its first argument.

Assumption A2 means that the problem is equivalent to one in which the benefit function is a smooth, quasiconcave function of a single state variable.

The crucial assumption, which differentiates the model from earlier literature, is that the cost of adjusting $x(\cdot)$ may depend on time or/and on the level of x . We now consider the former case; the latter is similar and is discussed below.

To make matters as simple as possible, we divide time into periods and assume that the cost of adjustment can take on only two values: high, c_h , and low, c_l . The cost is equal to the lower value for adjustment at the beginning of a period, and to the higher value for adjustments within a period. Some notation will be helpful. Let $\mathfrak{S} \equiv \{\tau_0, \tau_1, \dots\}$ consist of the beginnings of each period. The interval $[\tau_i, \tau_{i+1})$, $i=1, 2, \dots$ will be called *period i*. Whenever the adjustment takes place at $t \in \mathfrak{S}$, its cost is c_l . Such adjustment will be called regular adjustment and the *incidence of regular*

¹⁰ A somewhat stronger restriction is that all but one (say, the first) of the elements of the vector of state variables $\bar{x}(t)$ are fixed, i.e. $\bar{x}(t) \equiv (x(t), x_2^0, x_3^0, \dots, x_k^0)$.

adjustments (*IRA*) will be the proportion of all adjustments which are regular, $0 \leq IRA \leq 1$.

A3. The cost of adjustment is:

$$c(t) = c_h + I(t) \cdot (c_l - c_h), \quad c_h \geq c_l \quad (1)$$

where $I(t)$ is an indicator function, given by:

$$I(t) = \begin{cases} 1 & \text{for } t \in \mathfrak{S} \\ 0 & \text{for } t \notin \mathfrak{S} \end{cases} \quad (2)$$

As the focus of the paper is regular behaviour, we further assume that periods are of the same length, i.e. τ_i 's are evenly spaced over time:

$$\tau_i = \tau_0 + n \cdot \tau, \quad n = 1, 2, \dots \quad (3)$$

Obviously, the larger is the difference between the high and low values of costs, the more tempting is regular adjustment and so a large value of $c_h - c_l$ makes the problem trivial. Therefore we are careful not to make any assumptions about the size of the difference. All results hold even if the $c_h - c_l$ is arbitrarily small.

In this paper we concentrate on the simple nonstochastic case. In particular:

A4. The state variable $x(t)$ is assumed to change over time at a constant rate:¹¹

$$x(t) = x(t_0) \cdot e^{-\alpha(t-t_0)} \quad (4)$$

Without loss of generality, we assume $\alpha > 0$.

At the time of the first adjustment the policymaker's goal is to pick the sequences of times of adjustment and the new values of the state variable, $W \equiv \{x_0, (t_1, x_1), (t_2, x_2), \dots\}$ so as to maximize the present value of the benefits:

$$\text{maximize}_{\{t_i\}_{i=0}^{\infty}, \{x_i\}_{i=0}^{\infty}} PV(W) = \sum_{i=0}^{\infty} \left\{ \int_{t_i}^{t_{i+1}} B[x_i e^{-\alpha(t-t_i)}, \bar{y}(t), \bar{a}] e^{-\rho t} dt - c(t) e^{-\rho t_{i+1}} \right\} \quad (5)$$

¹¹ As already mentioned, an equivalent problem is when the optimal value of the state variable changes over time and adjustments are needed to keep the actual value close to the optimal value. The second application of our theory we consider in this paper, i.e. level-regular adjustment, falls into that category: The state variable in this case is the nominal price whose optimal value (the optimal real price) drifts over time. Optimal adjustment entails resetting the price to these drifting levels or, given the heterogeneous adjustment costs across levels, to a level with lower adjustment cost. This problem can be converted into the time-dependent problem by normalizing the drifting optimal value by its trend.

where $PV(W)$ denotes the present value of policy W , t_0 is the time of the first adjustment, ρ is the discount factor, and the first adjustment is assumed to be costless.¹²

The solution strategy we adopt is to start with the baseline case when \bar{y}, \bar{a} do not change over time and the cost of adjustment is constant and equal to its higher value, i.e. $c_l = c_h$. We then compare outcomes under heterogeneous costs with the baseline case. Note that in both cases the value of c_h is the same; they differ by the value of c_l . To set notation, the optimal policy under either case will be denoted with a “*” and the policy under the baseline case will be denoted with a “^”.

Lemma 1.

Assume $c_l = c_h$. Let $\hat{W}^* \equiv \{(\hat{x}_0^*, (\hat{t}_1^*, \hat{x}_1^*), (\hat{t}_2^*, \hat{x}_2^*), \dots)\}$ denote the optimal policy, and $\hat{T}^* = \{\hat{t}_1^*, \hat{t}_2^*, \dots\}$ denote the set of the optimal adjustment times. Then \hat{W}^* is recursive: $\forall i: \hat{x}_i^* = x^*$ and, for all i $\hat{t}_{i+1}^* = \hat{t}_i^* + \Delta \hat{t}^*$. Also, \hat{W}^* is unique.¹³ Finally, $\hat{B}(\hat{x}, \bar{y}, \bar{a}) - \hat{B}(\hat{x}e^{-\rho \Delta \hat{t}}, \bar{y}, \bar{a}) = \rho c_h$

The proof is essentially the same as in Sheshinski and Weiss (1977).

2.1. Positive Incidence of Regular Adjustments.

We now turn to showing an existence result: except in unlikely circumstances, the incidence of regular adjustments is positive: $IRA > 0$. In other words, it is optimal for the policymaker to take advantage of the lower adjustment costs. Of course it is important that the incidence of regular adjustment is not driven by the cost saving. Proposition 1 below shows sufficient conditions under which, when $c_l < c_h$, we get $IRA > 0$ even if the difference $c_h - c_l$ is arbitrarily small. The proof is based on the following approximation of real numbers with rational numbers:

Lemma 2.

For every $x, K > 0$ there exist integers N_1, N_2 such that $N_2 \leq K$ and $|N_2 \cdot x - N_1| < 1/K$.

Proof: see Niven (1961).

¹² As we consider the nonstochastic case here, we omitted expectations from equation (5).

¹³ Note that, since the optimized present value of benefits may be negative, no additional restrictions are placed on the values of the parameters and the momentary benefit function B .

The lemma can be applied to the problem considered here by setting $x = \Delta t^* / \tau$. It implies that, if the policymaker follows a policy of adjusting once every Δt^* (which is optimal when costs of adjustment are constant), eventually an adjustment will take place arbitrarily close to the beginning of a period. Given the notation, for an arbitrary value of K , the N_2 th adjustment will be within $1/K$ of the beginning of period N_1 .

Since the N_2 th adjustment is close to the beginning of a period, the firm needs to alter its timing just a little to take advantage of the lower beginning-of-period adjustment costs. It will do so as long as the reduction in adjustment costs exceeds the loss in benefits. Obviously, as already mentioned, we do not want the result to depend on the difference $c_h - c_l$. A sufficient condition for the results to hold regardless of the size of $c_h - c_l$ is that the slope of the benefit function be bounded; this is the motivation for assumption (b) below:

Proposition 1.

Let $W^* \equiv \{x_0^*, (t_1^*, x_1^*), (t_2^*, x_2^*), \dots\}$ denote the optimal policy, and $T^* = \{t_0^*, t_1^*, t_2^*, \dots\}$ denote the set of the optimal adjustment times, when $c_l < c_h$.

Assume that:

- (a) $c(t)$ meets (1)-(3);
- (b) the time of the first adjustment $t_0 \in \mathfrak{S}$
- (c) for every \bar{y}, \bar{a} there exists $A < \infty$ such that, for every t , $|B'(x(t))| < A$;

Then $\{T^* \setminus \{t_0\}\} \cap \mathfrak{S} \neq \emptyset$.

Proof.

Without loss of generality let the time of the first adjustment be $t_0 = \tau_0$. The proof is by contradiction. Assume that $T^* \cap \mathfrak{S} = \{t_0\}$. Therefore, by Lemma 1, the set of optimal adjustment times is \hat{T}^* , with $\hat{t}_0^* = \tau_0$. By Lemma 2, setting $A=K$, there exist two positive integers N_1 and N_2 such that:

$$N_2 \Delta \hat{t}^* - N_1 \tau < (1/\rho) \ln(c_h / c_l) \quad (6a)$$

$$N_2 \Delta \hat{t}^* - N_1 \tau < \rho(c_h - c_l)/(2A) \quad (6b)$$

When (6a) and (6b) are met we have:

$$PV(\hat{W}^*) < PV(W) \leq PV(W^*)$$

Proposition 1 shows that, when the adjustment costs vary over time as postulated in A3 and the first adjustment is at the beginning of period 0 ($t_0 \in \mathfrak{S}$), under general conditions the policymaker would, sooner or later, take advantage of the lower costs of adjustment. Assumption (c) requires a discussion. If the time of the first adjustment $t_0 \notin \mathfrak{S}$, it is possible that the policymaker will never take advantage of lower adjustment costs. This would be the case if, for example, $\Delta \hat{t} = \tau$ (i.e. when the optimal time between adjustments under constant costs is equal to the length of a period) and the difference between c_l and c_h is small.

In many environments, however, $t_0 \notin \mathfrak{S}$ is an unlikely outcome. This is because the timing of the whole sequence of subsequent adjustment times, $T^* \setminus \{t_0\}$, often depends on the time of the first adjustment. For example, the timing of subsequent visits to a doctor is set relative to the initial visit, dates of subsequent delivery depend on initial delivery etc.¹⁵ From now on we will assume that $t_0 \in \mathfrak{S}$.

By Proposition 1, at least one time of adjustment under W^* coincides with the beginning of a period. To set notation, assume that the first such adjustment is the N th adjustment, and it takes place at the end of period k . Denote such a policy as $W_{N,k}^*$.

This means that, under $W_{N,k}^*$, $t_N^* = \inf \{T^* \setminus \{t_0\} \cap \mathfrak{S}\} = \tau_k$,

It is easy to see that, for a given benefit function and adjustment costs, the optimal policy need not be unique. It is possible that $\tau_k < \hat{t}_N^* < \tau_{k+1}$ and $PV(W_{N,k}^*) = PV(W_{N,k+1}^*)$, i.e. the policymaker is indifferent between accelerating or delaying the N th adjustment.

The analysis of multiple equilibria in the current framework is complex. We therefore assume that, if $PV(W_{N,k}^*) = PV(W_{N,k+1}^*)$ then $W^* = W_{N,k}^*$, i.e. whenever two policies yield the same present value of benefits, the policymaker chooses the policy with later adjustments.

¹⁵ In environments in which the timing of adjustment is dictated by custom this need not be the case. For example a clothing store which opens in June may not be willing to have a sale shortly after the opening.

Proposition 2.

W^* is recursive:

$$W^* = \left\{ \left[(\tau_0, x_0^*), (t_1^*, x_1^*), \dots, (t_{N-1}^*, x_{N-1}^*) \right], \left[(\tau_k, x_k^*), (t_{N+1}^*, x_{N+1}^*) \dots, (t_{2N-1}^*, x_{2N-1}^*) \right], \dots \right\}$$

Proof.

W^* can be written as: $W^* = \left\{ \left[(\tau_0, x_0^*), (t_1^*, x_1^*), \dots, (t_{N-1}^*, x_{N-1}^*) \right], W^*(\tau_k^+) \right\}$, where $W(\tau_k^+)$ is the remainder of the optimal policy from period τ_k forward. Since W^* is optimal and unique, by the principle of optimality $W(\tau_k^+)$ is the solution to the problem of maximizing the present value of the benefits, starting in period τ_k . But this problem is identical to the original problem, as can be checked by substituting, $t_i^* = t_{i-N}^*$. Therefore, $t_{2N}^* = \tau_{2k}$ and for every i such that $N < i < 2N : t_i^* \notin \mathfrak{S}$. The proposition follows by induction. ■

The crucial question arising in this framework is the empirical incidence of adjustment at times in \mathfrak{S} , i.e. the value of IRA . By proposition 2, $IRA=1/N$: as the first adjustment in \mathfrak{S} is the N th adjustment and the optimal policy W^* is recursive, every $1/N$ th adjustment is in \mathfrak{S} .¹⁶

Proposition 1 is an existence result: it shows that $IRA>0$ as long as $c_h > c_l$ (even if the difference $c_h - c_l$ is arbitrarily small) and the benefit function is not too steep, and subject to the discussion above. While this result is interesting, it has little empirical content, especially given the fact that the starting point of the analysis is the observation that many policies are, indeed, regular: some prices are changed at the beginning of the year, firms sometimes order a delivery of multiple truckloads etc. Therefore we now turn to the analysis of the factors which determine the incidence of regular adjustment.

2.2. Factors Affecting the Incidence of Regular Adjustment (IRA).

We address the determination of IRA in two steps. First, we consider the determinants of the incidence of regular adjustment for a single policymaker. Then we

¹⁶ Of particular interest is the special case of $IRA=1$, i.e. when $N=1$ $T^* \subseteq \mathfrak{S}$ and the firm never pays c_h . Of course, T^* may be a proper subset of \mathfrak{S} (i.e. $T^* \subset \mathfrak{S}$) when $N=1$, for example if the optimal adjustment frequency is once every two periods.

analyze empirical predictions of the model, under two alternative assumptions regarding the differences between policymakers.

Before we proceed we need to define precisely when a policymaker will deviate from the optimal policy \hat{W}^* (i.e. the policy that she would have followed if adjustment costs were constant) to take advantage of the lower costs. We call it the *shift range*:

Definition: The shift range $S_i = [\hat{t}_i^* - a_i, \hat{t}_i^* + b_i]$ is an interval such that the following two conditions are met:

- (a) the policymaker moves the i th adjustment from \hat{t}_i^* to some τ_j if and only if $\hat{t}_i^* - a_i \leq \tau_j < \hat{t}_i^*$;
- (b) the policymaker moves the i th adjustment from \hat{t}_i^* to τ_{j+1} , if and only if $\hat{t}_i^* < \tau_{j+1} \leq \hat{t}_i^* + b_i$.

In other words the policymaker moves the timing of the i th adjustment, which falls within period j , to the beginning of period j or to the beginning of period $j+1$ if and only if the optimal timing under constant adjustment costs falls in the shift range S_j . Due to the fact that, by Proposition 2, W^* is recursive, the index i is counted from τ_0 (or, equivalently, from the last time adjustment is at the beginning of a period). As before, we assume that if the policymaker is indifferent between accelerating or delaying adjustment, she chooses to delay it.

We now make two additional simplifying assumptions that are sufficient, although not necessary, to derive the remaining results:

A5. The benefit function is quadratic in the state variable x :

$$B[x(t), \bar{y}(t), \bar{a}] = \Phi\left\{\left(-qx^2 + rx + s\right), b[\bar{y}(t), \bar{a}]\right\}$$

where the functional $\Phi(\cdot, \cdot)$ is an identity in its first argument.¹⁷

A6. The discount factor $\rho = 0$.

The shift range S_i determines the willingness of the policymaker to take advantage of the lower adjustment costs. The size of S_i depends on two factors: the size of the difference $c_h - c_l$ and the value of benefits foregone by departing from \hat{W}^* .

¹⁷ This formulation allows for different effects on the value of benefits of other state variables and of parameters, for example multiplicative $B[x(t), \bar{y}(t), \bar{a}] = \left(-qx^2 + rx + s\right) \cdot b[\bar{y}(t), \bar{a}]$ or exponential

$B[x(t), \bar{y}(t), \bar{a}] = \left(-qx^2 + rx + s\right)^{b[\bar{y}(t), \bar{a}]}$

The policymaker faces a trade-off between reducing adjustment cost and the reduction in benefits brought about by not following \hat{W}^* . The loss depends on how fast benefits decline as the time of adjustment varies. This, in turn, depends on the slope of the benefit function. A benefit function that is, at a given distance from its maximum, flat, makes the loss small and so the policymaker is willing to vary adjustment time to save on adjustment cost.

Proposition 3:

Let B^1 and B^2 be two benefit functions with parameters q^1 and q^2 and S_i^1, S_i^2 be their respective shift ranges. If $q^1 > q^2$ then, for all i , $b_i^1 \leq b_i^2$ and $a_i^1 \leq a_i^2$.

Proof.

We consider the postponement of the times of the i th adjustments \hat{t}_i^{*1} and \hat{t}_i^{*2} , i.e. that $b_i^1 \leq b_i^2$; the proof for the acceleration of \hat{t}_i^{*1} and \hat{t}_i^{*2} is analogous. Assume i is the lowest index such that $\hat{t}_i^{*1} \in S_i^1$. This means \hat{t}_i^{*1} is delayed until the nearest beginning of the period, say period $k1$: $t_i^{*1} = \tau_{k1}$ and all prior adjustments are within periods. It is easy to show that, since the discount rate is zero by A6, the times between adjustments are all of equal length: $t_j^1 - \tau_0 = (\tau_k - \tau_0) \cdot (j/i)$ for all $j \leq i$. Therefore shifting the time of the i th adjustment from t_i^{*1} to τ_{k1} involves extending all i times between adjustments by $(\tau_{k1} - t_i^{*1})/i$. Since $\hat{t}_i^{*1} \in S_i^1$, the saving on adjustment costs, $c_h - c_l$ is greater than i times the loss of extending adjustment time (and changing appropriately the new value of x).

Assume now that $\tau_{k2} - \hat{t}_i^{*2} = \tau_{k1} - \hat{t}_i^{*1}$ where τ_{k2} is the first beginning of the period following \hat{t}_i^{*2} . The benefit from postponing \hat{t}_i^{*2} is the saving on adjustment costs and is the same as for B^1 but, as $q^1 > q^2$, the cost of the postponement is lower. This means that, for B^2 , the benefit exceeds the cost. Therefore $t_k^{*2} = \tau_m$, $m \geq k2$, which implies $b_i^1 \leq b_i^2$. ■

2.3. The Number Problem.

Proposition 3 shows that, for a given difference $\hat{t}_i^* - \tau_k$ and $\tau_{k+1} - \hat{t}_i^*$, the flatter is the benefit function at the optimal choice, the more likely is the policymaker, *ceteris paribus*, to take advantage of lower adjustment costs. But that does not mean

that the relationship between the second derivative of the benefit function and the incidence of regular adjustment is unambiguous. This is because the differences $\hat{t}_i^* - \tau_k$ and $\tau_{k+1} - \hat{t}_i^*$, depend on the parameters of the model in a way that depends crucially on what we call the *number problem*. Essentially, when $c_l = c_h$ for any benefit function the optimal time of adjustment may happen to fall close to the beginning of a period and so a high incidence of regular adjustment may happen just by coincidence.

To provide an example, consider a given problem in which $t_0 = \tau_0$ and $\Delta \hat{t}^*$ is a well-defined, continuous function of the exogenous variables \bar{y} and the parameter vector \bar{a} . Assume further that, for some particular values of the exogenous variables and parameters, \bar{y}_0 and \bar{a}_0 , we have $\Delta \hat{t}^* = \tau$, i.e. under constant adjustment costs it is optimal for the policymaker to always adjust at the beginning of the period. In this case the policy is completely regular ($IRA=1$) in a neighborhood of (\bar{y}_0, \bar{a}_0) but $IRA < 1$ outside this neighborhood. Since there is, in general, nothing special about (\bar{y}_0, \bar{a}_0) , the resulting policy is regular just by coincidence.

As a more specific example, assume that $B=B(x,a)$, i.e. the benefit function depends on the state variable and one parameter. Assume that the parameter is observable and its value is positively related to $\Delta \hat{t}^*$. This is the setup considered by Sheshinski and Weiss (1977), where $B[.]$ is the real profit function of a monopolist, x is the real price and a is the inflation rate. Let adjustment costs vary as postulated here. Assume that a researcher studies six policymakers and the observable parameter a is distributed across policymakers in such a way that their (unobservable) optimal periods of adjustment under constant cost, $\Delta \hat{t}_i^*$, are equal $10+i/32$ months, $i=15, \dots, 20$. Assume further that the difference between the high and low level of adjustment costs is so small that they never depart from \hat{W}^* . The incidence of regular adjustments she observes is summarized in Table 1 below:

Table 1

Monthly frequency of adjustment (%)	9.41	9.44	9.47	9.50	9.52	9.55
Incidence of regular adjustment	0.13	0.03	0.06	0.03	0.50	0.03

There is no easy way around the number problem. A potential solution is suggested by the empirical implementation below, which treats the average frequency of adjustments as an indicator of cross-policymaker heterogeneity. If the average frequency of adjustment is the variable of interest, the number problem is eliminated if the following condition on the empirical distribution of $\Delta \hat{t}^*$ over time is met:

C1. The empirical distribution of $\Delta \hat{t}^*$ on $\{\tau_{i-1}, \tau_i\}$ is independent of i .

Under this condition, the probability of finding a policymaker for whom the timing of the k th adjustment, $k\Delta \hat{t}^*$, is within a given distance from the beginning of the period is the same for all periods.

The problem with this condition is that it is not met in practice due to truncation of the range of $k\Delta \hat{t}^*$ both from below and above. The truncation from below is due to the fact that, under lump-sum costs, $\Delta \hat{t}^*$ is bounded away from zero but $\Delta \hat{t}^*$ is not bounded away from above from $\tau, 2\tau, \dots$. The truncation from above is due to the fact that the limited length of the sample makes it impossible to observe policies $W_{N,k}^*$ for which $k\tau$ exceeds the length of the sample. Therefore it is possible for results of empirical tests of the model to be dominated by the number problem. This makes it difficult to interpret rejections of the model since an empirical test of the model is a joint test of the relationship between benefit function shape and the incidence of regular adjustments as well as the fact that the number problem is “averaged out” in the data set. But the number problem is essentially a statistical issue unlikely to be affected by the considerations of the model. Hence it becomes irrelevant if the results of empirical tests are consistent with the model.

2.4. Empirical Predictions under Different Assumptions about Policymaker Heterogeneity.

The discussion above indicates that a model in which all firms are identical and their adjustment costs vary as postulated in A3 does not, in general, have unambiguous empirical implications. To obtain empirical predictions of the model, and avoid results being dominated by the number problem, heterogeneity across

policymakers is needed. Furthermore, testing should use a large data set. The second requirement rules out, for practical purposes, time-series analysis since long data series on the timing of adjustments are difficult to obtain. In the next subsection we therefore discuss empirical implications of the model under two different assumptions on cross-sectional heterogeneity across policymakers.

To obtain empirical predictions of the model we consider alternative sources of differences across policymakers: (a) with respect to the shape of the benefit function, (b) with respect to the value of adjustment costs and (c) with respect of the rate of deterioration, α , of the state variable. We focus on the first two as they are tested in the next section; our data are insufficient to test model implications for the third one.

In terms of the model the benefit function heterogeneity is represented by the value of the parameter q , which determines the concavity of benefit function,. The adjustment cost heterogeneity is represented by the high value of the adjustment cost, c_h , with the difference $c_h - c_l$ kept constant.

Both types of heterogeneity have been used in the modeling of optimal pricing policies under the assumption of costly price adjustment. The first type was considered by Konieczny and Skrzypacz (2006) who analyze an equilibrium optimal pricing model with costly price adjustment and consumer search for the best price. Their model, briefly described in the next section, implies that the greater is the consumer propensity to search for the best price in a given market, the greater is the value of the parameter q . The second type of heterogeneity was considered by Dotsey, King and Wolman (1999) who develop a tractable framework incorporating costly price adjustment into a general equilibrium model. In their approach firms differ with respect to their adjustment costs.

As shown below, the two assumptions produce opposite results and so an empirical study we propose can, potentially, discriminate between them under the joint hypothesis that adjustment costs vary as postulated in our model.

Proposition 4.

Consider an environment with many policymakers whose benefit functions are as in A5, and whose adjustment costs vary over time (or over states) as postulated in A3. For all policymakers let $\hat{t}^* \in (0, \tau_n]$, $n \geq 1$ and assume that

the condition C1 is met for all $i \leq n$. Assume further that policymakers are identical except for one source of heterogeneity across policymakers:

(a) If the differences across policymakers are due to differences in the value of q , then the lower is q , the less frequent is adjustment and the higher is the incidence of regular adjustment.

(b) If the differences across policymakers are due to differences in the value of c_h and c_l (so that $c_h - c_l$ is the same across policymakers) then the higher c_h , the less frequent is adjustment and the lower is the incidence of regular adjustment.¹⁸

(c) If the differences across policymakers are due to differences in the value of α , then the lower is the value of α , the less frequent is adjustment and the higher is the incidence of regular adjustment.

Proof:

The effect on the frequency of adjustment in 4(a) follows directly from the Lemma (Konieczny and Skrzypacz (2006)); in 4(b) it follows directly from Sheshinski and Weiss (1977), section 5 and in 4(c) it follows directly from Proposition 2 in Sheshinski and Weiss (1977) since the quadratic benefit function meets their condition (M).¹⁹ The effects on *IRA* follow directly from proposition 3.

■

III. Empirical Evidence.

We test the model by analyzing optimal pricing policies at the firm level. In the pricing application the benefit function $B[.,.,.]$ is the profit function of a monopolistic, or monopolistically competitive firm which produces a single product. Under general inflation at the rate α , its real price falls over time. To reset it the firm changes its nominal price, which involves paying a lump-sum *menu* cost.

¹⁸ To avoid confusion note that in (b) there are two sources of heterogeneity in adjustment costs. The first source is heterogeneity in the size of adjustment costs over time (or over states), as postulated in A3. It is the same for all policymakers. The second source is heterogeneity across policymakers. In (a) and (c) the differences in adjustment costs are due to the first source only.

¹⁹ As long as $\hat{x}^* \exp(-\alpha \Delta \hat{t}^*) > \tilde{x}^* / 2$, where \tilde{x}^* is the benefit-maximizing value of x in the absence of any adjustment costs. This inequality is met in our empirical study in the next section.

3.1. Tested Hypotheses.

The data allow us to analyze the incidence of both time-regular and state-regular policies. We define a time-regular policy as price adjustment at the beginning of the year, and, separately, as price adjustment at the beginning of a quarter. We will refer to such policies as *seasonal* price setting. State-regular policies involve choosing *attractive* prices: prices that end in a nine or round prices. The definition (values) of attractive prices is given in the Appendix.

Our data set, which we describe below, does not allow for a direct test of Proposition 4(c) as the variation in the inflation rate in the data is small. Therefore we concentrate on the differences across firms in the shape of the profit function and in the values of adjustment costs.

Our H_0 hypothesis, implied by Proposition 4(a), is that the adjustment costs vary as postulated and that the differences across policymakers are due to heterogeneity in the shape of the profit function, as in Konieczny and Skrzypacz (2006). The alternative, implied by Proposition 4(b), is that the differences are due to heterogeneity in the level of price adjustment costs, as in Dotsey, King and Wolman (1999).

The data set used to test the model is extensive and the variation in the endogenous variable is large. Therefore we would treat an insignificant estimated coefficient on the adjustment frequency as a rejection of the model, notwithstanding the number problem. If the coefficient is negative and significant, we treat it as support for the joint hypothesis that menu costs vary as postulated and heterogeneity across markets is due to differences in the shape (curvature) of the profit function. If it is positive and significant, we treat it as support for the joint hypothesis that menu costs vary as postulated and heterogeneity across markets is due to differences in the size of the menu costs.

Since neither the curvature nor the value of adjustment costs is observable in our data, a direct test of the model is not possible. However, an indirect test of the model can be performed with another variable acting as an instrument for the unobservable variable. In view of Proposition 4, we treat the adjustment frequency as the instrument.

Before we turn to the data, we now briefly describe the two underlying models of policymaker heterogeneity.

Konieczny and Skrzypacz (2006) analyze a model, based on Bénabou (1992), in which firms face nominal adjustment costs and consumer search for the best price. They consider a market for a single good which is supplied by a continuum of firms, each with the same marginal cost MC . Firms set nominal prices so as to maximize the average value of real profits per unit of time. Nominal prices are eroded by constant inflation at the rate α . As price adjustment is costly, nominal prices are changed infrequently. In the absence of perfect synchronization prices differ across firms.

Each period a new cohort of v consumers per firm arrives in the market. Each consumer buys 0 or k units of the good and exits the market. Consumers search for the best price. They are heterogeneous in terms of their adjustment costs c , which is distributed uniformly over the range $[0, C]$ in each cohort. Heterogeneity across markets is due to differences in the values of the parameters k and C , which determine the propensity to search for the best price, and the density of customers, v .

The model is directly applicable to our framework. Konieczny and Skrzypacz (2006) show that the profit function is, using our notation²⁰ $B(x) = -qx^2 + rx + s$. The parameter q , which is crucial in our study since it determines the concavity of the benefit function, is a simple function of k , C and v : $q = vk^2/C$. More active search for the best price, due to a large amount spent on the good (large k) or low search costs (represented by a low maximum value C), or a large number of customers (large v), lead to more concave profit functions.

Dotsey, King and Wolman (1999) develop a general equilibrium framework for the dynamic analysis of the effects of various macro disturbances in the presence of price adjustment costs. In their model both firms and consumers are long lived. Consumers have Dixit-Stiglitz preferences for variety and so firms are monopolistically competitive. Heterogeneity is due to differences in the value of adjustment costs: firms draw them independently over time from a continuous distribution.

The model is a general equilibrium one, but for fixed values of exogenous parameters it can be interpreted as a multi-market model. As the Dixit-Stiglitz preferences imply constant-elasticity demand, the profit functions are not quadratic. The results of our model hold when the profit functions can be approximated with a quadratic, i.e. for low values of adjustment costs and/or low inflation. While the

²⁰ Here x is the real price, $q = vk^2/C$, $r = q(C/k + E(x) + MC)$, $s = MC \cdot [C/k + E(x)]$ and $E(x)$ is the average price in the market.

values of adjustment costs are not observable, during the period under study the inflation rate in Austria never exceeded 3.4%.

3.2. Data.

To test the model we use a very large Austrian data set. It is the data set analyzed in Baumgartner et al. (2005) who studied the stylized facts of price setting in Austria.²¹ It contains monthly price quotes collected by the Austrian statistical office, which are used in the computation of the Austrian CPI. The sample spans the period from January 1996 to December 2003 (96 months) and contains about 40 000 elementary price records per month. Overall, the data set contains about 3.6 million individual price quotes and covers roughly 80% of the total Austrian CPI. Each record includes, in addition to the nominal price, the information on the product category, date, outlet (shop), packaging type, a sales indicator and a number of other indicators.

We identify a “policymaker” with a product category, i.e. a product at the elementary level included in the CPI basket, for example milk, rather than an individual store/product pair.²² We need a large number of price changes to compute the conditional frequencies used in the empirical testing. Thus we implicitly assume that heterogeneity is across markets and all firms operating on a given market (selling a given product) share the same profit function or adjustment costs.

The original data set (used in Baumgartner et al., 2005) contains a total of 668 product categories. We excluded 151 product categories with administered prices, excessive price changes and products for which we had data for several varieties.²³ We eliminated all products with an average size of price changes of more than 50%. We suspect that, in such cases the definition of the product (on which no direct information is available in the data set) has been changed during the sample period. Hence the requirement of Proposition 4(c) – see footnote 19 – is met in our data. This leaves 517 product categories for our analysis. The average frequency of price

²¹ They describe in detail the data and some manipulations which have been carried out prior to the statistical analysis.

²² Treating an individual store/product pair as a policymaker would require calculating the average frequency of price changes from too few observations, especially for stores which change prices infrequently.

²³ For some product categories the data set contains prices for several varieties (for example prices of car insurance for different types of cars). These prices are usually changed jointly and so, in such cases, we considered only the price for the variety with the highest CPI weight.

changes is between 0.8% per month (chipboard screws) and 91% per month (package holidays).

3.3. Causality.

The analysis raises the issue of causality. Our model implies that infrequent price changes and high incidence of regular adjustments coincide because of a common causal characteristic (flat profit function or low adjustment costs). But several studies in the Inflation Persistence Network imply causation from what we call regular adjustment to the frequency of price changes. In the data set we are using, Baumgartner et al (2005) find that the probability of price adjustment, conditional on the last price being an attractive price, is lower than the unconditional probability. Similar results have been documented by Álvarez and Hernando (2004) for Spain, Aucremanne and Dhyne (2005) for Belgium, Veronese et al. (2005) for Italy, Lünemann and Mathä (2005) for Luxembourg, Hoffmann and Kurz-Kim (2005) for Germany and Dhyne et al. (2005) for a panel of euro area countries. This means that, were we simply to analyze the relationship between the frequency of price changes and the incidence of attractive prices, we may discover a negative relationship where causality goes from the proportion of attractive prices to low price changing frequency: in markets in which the proportion of attractive prices is high, the average frequency of price changes will be low.

In order to overcome this potential problem of reverse causality in our regression, we need a measure for the frequency of price changes that is independent of the proportion of attractive prices. Therefore we condition the frequency of adjustment on, separately, attractive and non-attractive prices: for product category *i* we calculate the average conditional frequency of a price change given that the last price is an attractive price and, separately, as the conditional frequency of price changes given that the last price is not an attractive price. We then use both conditional frequencies in the regression as explanatory variables. The use of both conditional frequencies avoids the results being affected by the mixture of attractive and other prices in the given market.

While we are not aware of similar empirical evidence for seasonal price setting, we suppose the same is true in that case as well: the probability of price adjustment conditional on the previous adjustment taking place at the beginning of the year (or quarter) would be lower than the unconditional probability of adjustment.

Therefore, we adopt the same approach in the regressions explaining the incidence of seasonal price setting using, as explanatory variables, both the conditional frequency of price changes if the last price change was at the beginning of a year/quarter and the conditional frequency if it was within a year/quarter.

3.4. Result for Time-Regular Adjustment.

We first discuss the results for time-regular policies. We implement the model by looking at the determinants of the excess proportion of price changes taking place at the beginning of a year and, separately, at the beginning of a quarter. We call such adjustments *seasonal*. Empirically, price changes in the Austrian data are, indeed, more frequent at the beginning of the year and, for some products, also at the beginning of a quarter (see Baumgartner et al., 2005).

We estimate the following equation:

$$Seas_i = f\left(F_i^{seas}, F_i^{nseas}, \bar{z}_i\right) \quad (8)$$

where i indexes markets (product categories), $Seas_i$ is the excess proportion of price changes at the beginning of a period (a year or a quarter), defined below, F_i^{seas} is the average frequency of price changes in market i conditional on the previous price change having taken place at the beginning of a period, F_i^{nseas} is the average frequency conditional on the previous price change having taken place within a period and \bar{z}_i is a vector of control variables.

According to Proposition 3, firms which have a flatter profit function at \hat{t}^* will change their prices less frequently, by a larger amount and prefer a seasonal pattern of their price adjustment, i.e. have a larger proportion of price changes at the beginning of a year or a quarter. Thus under H_0 (i.e. when firms differ in terms of the concavity of the profit function), the share of price changes at the beginning of a period should be negatively related to the (conditional) frequency of price changes and positively related to the average size of adjustment. Under H_1 (i.e. when firms differ in terms of menu costs) the share of price changes at the beginning of period should be positively related to the (conditional) frequency of price changes and negatively related to the average size of adjustment.

The dependent variable in regression (8) is the ratio of the number of price changes taking place at the beginning of the period to the number of all price changes in that period, normalized to avoid it being bounded. Given that our data are monthly we adopt two definitions of a period: a year and a quarter. In yearly regressions we compute the ratio of the number of price changes in a January of any year to all price changes in the sample; in quarterly regressions we compute the ratio of the number of price changes in any January, April, July or September to the number of all changes in the sample. The (normalized) dependent variable is obtained by dividing the yearly (quarterly) statistics by the share of valid price observations at the beginning of the year (quarter). According to this definition, a number above 1 indicates that relatively more prices are changed at the beginning of the period than average.

The remaining control variables include the size of price changes, the average and the standard deviation of inflation for the product i , the degree of synchronization of price changes, the share of sales prices and dummies for broad good categories.

Under H_0 large price changes characterize firms with flat profit functions which, by Proposition 4(a), prefer seasonal adjustment while, under H_1 , large price changes characterize firms with high adjustment costs which, by Proposition 4(b), rarely adjust at the beginning of the period. Hence we expect the coefficient on adjustment size to be positive under H_0 and negative under H_1 .

The average inflation rate may matter since, *ceteris paribus*, the higher is inflation the more frequent and larger are price changes and the steeper is the profit function at \hat{t}^* . But the effect of inflation is indirect, operating through its impact on adjustment size (and frequency). Since we are controlling for the size of price changes in the regression, the coefficient on the average inflation rate represents the effect on seasonal adjustment holding constant the size of price changes. The model makes no predictions about this conditional effect. Another consideration, not addressed directly by our model, is the flexibility of optimal policy. Seasonal adjustment lowers the cost of adjustment but reduces the firm's pricing flexibility. Presumably, the higher is the inflation rate, the more important is flexibility and so we expect the coefficient to be negative. For the same reason we expect a negative effect of inflation variability, measured by the standard deviation of the monthly inflation rate for the product i . The

benefit of flexibility for state-contingent adjustment increases with the variability of the environment.

An important issue in analyzing the seasonal pattern of adjustment is that, in some industries, firms tend to change prices together. For example, clothing stores hold simultaneous sales. This tendency to synchronize price changes needs to be controlled for so as to avoid spurious correlation between seasonal patterns and the conditional frequencies of adjustment. Therefore we include, on the right hand side of the regression, the synchronization index of price changes as defined by Fisher and Konieczny (2000). It summarizes, with a single number, the tendency of prices to be changed together. The index is defined as the ratio of the sample standard deviation of the monthly proportion of price changes for a given product category to the standard deviation of the proportion under the assumption that price changes are perfectly synchronized.

The share of sales prices is included in the regression to control for situations where, in some markets, a large proportion of price changes are seasonal just because sales are held in January or at the beginning of a quarter. Similarly, some markets may be characterized by a low proportion of seasonal adjustment if sales are held within quarters.²⁴

Finally, we add dummy variables for broad good categories: processed food, energy, industrial goods and services (the reference category omitted in the regressions is unprocessed food) to account for fixed effects related to broad good categories. The probability of price changes differs significantly across these categories, and the differences are remarkably consistent across countries. In the eight comprehensive data sets (for Austria, Belgium, Finland, France, Luxemburg, Portugal, Spain and the U.S.)²⁵ as well as in the four smaller sets (for Poland, Germany, Holland and Italy)²⁶ the probability of price change is always the lowest for services, the highest for energy (except for Portugal, where prices of energy are

²⁴ Sales prices are identified by an indicator in the data set if a price is subject to a temporary promotion or a sale. In addition to these “flagged” sales, we identify “unflagged” sales as price reductions which are fully reversed in the following period. About 4% of all price observations in our data set are flagged sales prices and 1% are unflagged sales according to the above definition; for more information see Baumgartner et al., 2005.

²⁵ Baumgartner et al. (2005), Aucremanne and Dhyne (2004), Vilmunen and Laakkonen (2004), Baudry et al. (2004), Lünemann and Mathä (2005), Dias et al. (2004) Álvarez and Hernando (2005) and Bils and Klenow (2004), respectively.

²⁶ Konieczny and Skrzypacz (2006), Hoffmann and Kurz-Kim (2004), Jonker et al. (2004) and Veronese et al. (2005), respectively.

regulated and change relatively infrequently) and unprocessed food, followed by processed food and industrial goods.

The results of regression (8) are in Table 2.

Table 2.

Explaining the share of price changes at the beginning of a period (year, quarter)

Variable	Period = Year	Period = Quarter
Constant	0.591 ***	0.904 ***
Frequency conditional on seas. (F_i^{seas})	-2.926 ***	-1.771 ***
Frequency conditional on not seas. (F_i^{nseas})	-0.650	1.367 ***
Size of price changes	1.635	1.020 ***
Average inflation	0.592 ***	0.023
Standard deviation of inflation	-0.039	-0.007
Synchronization of price changes index	5.643 ***	0.676 ***
Share of sales prices	-1.371	-0.741 *
Processed food dummy	-0.117	0.011
Energy dummy	-0.293	0.102
Industrial goods dummy	-0.116	0.062 ***
Services dummy	0.552 ***	0.038
Number of observations	491	480
Adjusted R ²	0.458	0.221

Notes: Estimation method is OLS; standard errors are computed using White's correction for heteroskedasticity; inflation is calculated as monthly changes in the corresponding product category's sub-index; the number of products included is lower than the maximum 517 because some variables are not defined for all products; *** denotes significance at the 1%, ** at the 5% and * at the 10% level.

Table 2 shows the results for the period defined as a year (column 1) and the period defined as a quarter (column 2). Of the two specifications, price setting at the beginning of a year is empirically more relevant (the mean of the dependent variable is 2.01, indicating that price changes in January are 101% more frequent than in the other months of the year) than price adjustment at the beginning of a quarter (with a mean dependent variable of 1.16). Therefore, we regard the first column in the table as our standard specification and treat the results for price setting at the beginning of a quarter as an additional specification for a robustness check.

The crucial result is that the sign on both conditional frequencies: F_i^{seas} (i.e. if the last price change has taken place at the beginning of a year) and F_i^{nseas} (if the last price change has taken place within a year) is negative, as implied under H_0 . The

coefficient on F_i^{seas} is significant at the 1% level. In other words, in markets where prices are changed infrequently, a large proportion of these changes take place in January. The coefficient is both statistically and economically significant: increasing the conditional frequency if the last price change was at the beginning of the year F_i^{seas} by one standard deviation (17.3 percentage points) reduces the excess proportion of seasonal price adjustment by 0.29 standard deviations (0.51 in absolute terms). Note that in the regression we control for the synchronization of price setting (which is positive and significant) as well as for sales (which turn out not to be significant). While the Fisher-Konieczny index is not a perfect control²⁷, this reduces the likelihood that the negative sign is due to some markets being characterized by yearly price changes in January only, or by sales in January.

The coefficient on the size of price changes has a positive sign, as implied under H_0 but the effect is only marginally significant (at the 11% level). The coefficient on average inflation is positive; that on inflation volatility is, as expected, negative, but it is not significant.

Only services show a significantly higher share of price changes at the beginning of the year than the reference group (unprocessed food), which is related to the fact that many service prices in Austria are regularly changed in January (see Baumgartner et al., 2005). The commercial practice of sales and temporary promotions is obviously not an important determinant of seasonal price setting in January: the coefficient on the sales variable is negative but not significant. Finally, the coefficient on the synchronization variable is positive and significant at the 1% level. This indicates that in markets where firms synchronize price changes, adjustment in January is frequent.

The regression results for the quarterly pattern of adjustment, shown in the second column of Table 2, are similar to those in the first column with a few exceptions. The coefficient on F_i^{nseas} is positive and significant and the group effects are somewhat different. The coefficient on the size of price changes is positive, as expected under H_0 , and significant. As time-regular pricing is less pronounced than in yearly data it is not surprising that the adjusted R^2 is much lower than in the yearly regression. We conclude that the results for both regressions provide the same picture.

²⁷ It leaves several degrees of freedom as it summarizes, with just a single number, the monthly pattern in the proportion of price changes.

3.5. Results for State-Regular Adjustment.

We now turn to state-regular adjustment, i.e. adjustment under which the price charged is an attractive price. The empirical implementation of the testing requires a definition of attractive prices. There is no universal approach to defining attractive prices. We chose to adopt a broad definition that tries to capture all prices which are used by any firm or retailer as attractive prices. This comes at the risk of classifying too many prices as attractive. We think this is less problematic than missing important attractive prices. We require that the (percentage) differences between attractive prices be not affected by the order of magnitude of the prices (i.e. if 15.90 is an attractive price, so is 159 and 1590). This is important in our data set as it encompasses the replacement of the Schilling with the Euro, which involved the reduction of prices by roughly an order of magnitude (the exchange rate was 13.7603 Schillings/Euro). In addition, our definition is specifically tailored to the Austrian retail market as it takes account of the common pricing practices observed there (e.g. prices ending in 75 are not used as attractive prices in Austria). An explanation of the principles of our definition and (an excerpt of) a list of attractive prices are in the Appendix. With our definition, the average proportion of attractive prices in the data is 60.7%. It ranges from 0.07 for car insurance to 0.92 for digital cameras.

The estimated equation is:

$$Attr_i = f(F_i^{att}, F_i^{natt}, \bar{z}_i) \quad (9)$$

Under H_0 (i.e. when firms differ in terms of the concavity of the profit function), the share of attractive prices should be negatively related to the (conditional) frequency of price changes and positively related to the average size of adjustment. Under H_1 (i.e. when firms differ in terms of menu costs) the share of attractive prices should be positively related to the (conditional) frequency of price changes and negatively related to the average size of adjustment.

The share of attractive prices is a fractional response variable (bounded between 0 and 1), which implies that estimating a linear model is not appropriate. Therefore we transform the dependent variable to the log-odds ratio, $\ln(Attr_i/(1 - Attr_i))$ which is not bounded, and run an OLS regression on the transformed variable.²⁸ In order to obtain the marginal effect of each variable on the

²⁸ The log-odds model has been criticized for delivering marginal effects that may be inconsistent. An alternative approach used in Dhyne et al. (2005) is the quasi-maximum likelihood (QML) approach

dependent variable, the regression coefficients, β_k , have to be converted back by the formula $dy/dx = \beta_k \overline{Attr} \left(1 - \overline{Attr}\right)$ which usually is evaluated at the sample mean.

The control variables include the size of price changes, the average price level, the rate of inflation and its variability (measured by its standard deviation) and the share of sales prices. As before, the coefficient on the size of price changes is expected to be positive under H_0 and negative under H_1 . If attractive prices are more relevant at lower price levels (i.e. for cheaper goods), the average absolute price in a product category should be related negatively to the share of attractive prices. This variable also serves as a check if our definition of attractive prices is reasonable. For the reasons related to the flexibility of the optimal policy, outlined in the previous subsection, we expect the coefficients on the average product-specific inflation and on its variability to be negative. Finally, the incidence of attractive prices may be affected by temporary promotions and end-of-season sales; casual observation suggests that these prices are often attractive, and so we include the share of sales prices and promotions in each product category as another control variable in the regressions.

The results of regression (9) are in Table 3. We estimate the regression separately for the whole sample, and for the period prior to the introduction of the Euro.

The results are similar to those for the case of seasonal adjustment. The frequency of price changes (conditional on the last price being an attractive price, F_i^{att}) has a negative impact on the share of attractive prices, as implied under H_0 . This effect is significant at the 10% level for the whole sample, but is not significant for the short sample. The coefficient is economically significant: the marginal effect implies that, if the conditional frequency increases by one standard deviation (13.7 percentage points), the share of attractive prices is decreased by 0.44 standard deviations (10.2 percentage points).

proposed by Papke and Wooldridge (1996). It involves directly estimating a non-linear model of the explanatory variables and maximizing its likelihood function based on a Bernoulli distribution. We also performed estimations according to this approach, but the results (available upon request) are very similar.

Table 3.
Explaining the share of attractive prices.

Variable	Long Sample (96-03)	Schilling Sample (96-01)
Constant	0.231 ***	0.371 ***
Frequency conditional on attr (F_i^{att})	-0.745 *	-0.130
Frequency conditional on not attr (F_i^{natt})	0.649 *	-0.189
Size of price changes	0.622 ***	0.552 **
Average price (Schilling period)	0.000	0.000
Average price (Euro period)	0.000	
Average inflation	-0.102 **	-0.132 ***
Standard deviation of inflation	0.001	0.006
Share of sales prices	0.830 **	0.919 **
Processed food dummy	0.008	0.023
Energy dummy	-0.528 ***	-0.611 ***
Industrial goods dummy	-0.284 ***	-0.360 ***
Services dummy	-0.315 ***	-0.315 ***
Number of observations	505	507
Adjusted R ²	0.417	0.356

Notes: Estimation method is OLS on the log-odds ratio of the share of attractive prices; displayed coefficients are marginal effects of each variable on the share of attractive prices evaluated at the sample mean; standard errors are computed using White's correction for heteroskedasticity; inflation is calculated as monthly changes of the corresponding product category's sub-index; the number of products included is lower than the maximum 517 because some variables are not defined for all products; *** denotes significance at the 1%, ** at the 5% and * at the 10% level.

The effect of the frequency conditional on the last price not being attractive (F_i^{natt}), however, is positive and significant in the long sample. A possible explanation of this result is that firms have a strong incentive to follow an attractive pricing policy. For some reason they sometimes deviate from that policy and choose a price that is not attractive. But if they do so, they quickly return to an attractive price afterwards, which increases the conditional probability of a price change when the last price is not attractive.²⁹

The average (absolute) size of price changes in a market has a positive impact on the share of attractive prices in this market, as predicted implied under H_0 . The average price in the product category, which has been calculated separately for the

²⁹ That may be the case if the benefit from charging an attractive price is not lump sum, as modeled here, but is a stream of benefits. This is the implication of the rational inattention explanation of attractive prices by Basu (1997) and Bergen et al (2003). We leave such extension for future research.

Schilling period (1996-2001) and for the Euro period (2001-2003), does not affect the incidence of attractive prices. This result is reassuring as it indicates that the definition of attractive prices has been chosen appropriately. The average (monthly) inflation rate has a significant negative impact on the share of attractive prices while the volatility of inflation is not significant. Finally, the practice of sales and temporary promotions turns out to be an important additional determinant of attractive prices: the product categories with a higher share of sales and promotions are characterized by a higher share of attractive prices and the share of attractive prices is significantly lower for non-food items.

To check whether attractive price setting was not systematically different for Schilling and for Euro prices, in column 2 we show the regression results obtained for the sample period covered by our dataset when the Schilling was the legal tender in Austria (1996-2001).³⁰ Overall, the results for the short sample are qualitatively similar to the long sample. The exception is that the frequency of a price change, conditional on the last price not being attractive price has a negative sign and neither conditional frequency is significant.

To sum up, the regression results for both the time-regular adjustment and state-regular adjustment support H_0 : the joint hypothesis that adjustment costs vary as postulated and heterogeneity across markets is due to differences in terms of the concavity of the profit function, as suggested by Konieczny and Skrzypacz (2006). The results reject the joint hypothesis that adjustment costs vary as postulated and heterogeneity across markets is due to differences in the value of menu costs, as assumed by Dotsey, King and Wolman (1999).

IV. Conclusions and Extensions

Regular adjustment is ubiquitous in many environments, yet the reasons for such behaviour have not received much attention. In this paper we make a small step towards explaining the incidence of regular adjustment. It is attributed to the heterogeneity in adjustment costs across time/states and the heterogeneity in the shape of the benefit function across policymakers. The results show that our assumption on

³⁰ The sample period from the introduction of the Euro to the end of our sample (2002-2003) is too short to be analysed separately.

adjustment heterogeneity, despite its remarkable simplicity, is sufficient to account for the observed pricing behaviour.

Understanding the motives for regular adjustments is important for both theoretical and practical reasons. In the Fisher-Taylor and Calvo frameworks, and the vast macroeconomic literature based on them, time-regular adjustment is simply assumed. Models are developed and calibrated on the assumption that pricing policies are time contingent or, at best, that both time and state contingent policies are present but the division of price setters between the two types is fixed. While convenient, this assumption is unsatisfactory when macroeconomic conditions change. As an example, consider the empirical findings of Gagnon (2006) who studies pricing policies in Mexico. His data are unique in that they cover a wide range of inflation rates. He finds that changes in the inflation rate do not affect adjustment frequency when inflation is low, but do when inflation rate is high. This is consistent with implications of our model: under H_0 the lower is the inflation rate, the higher is the incidence of regular policies.

Ignoring the fact that the incidence of regular policies is an endogenous variable may lead to erroneous predictions. For example the effect of low inflation on the effectiveness of monetary policy depends on the source of the stability. If the reason why inflation has been low and stable in recent years is mostly due to monetary policy, then we can expect greater incidence of regular adjustments and increased monetary effectiveness. On the other hand, assume inflation is low because of greater competition. This raises demand elasticity and, so, by increasing the concavity of the profit functions, lowers the incidence of regular price adjustments and reduces the effectiveness of monetary policy.

In future work we plan to consider stochastic inflation. We expect the results would not change much as the optimal, state-contingent pricing policy under stochastic inflation is similar to the policy under a constant inflation rate. The analysis of stochastic inflation should bring out the benefits of flexibility. Policymakers who adopt regular adjustment reduce their flexibility. The understanding of the costs and benefits of flexibility is not only of intrinsic importance to these policymakers but is also important for more general considerations. Monetary policy is more effective when nominal price adjustments are regular.

One way of viewing state-contingent (as opposed to regular) adjustment is that it provides the option of flexibility, at the cost of raising adjustment costs. Assume

that there is a setup cost of switching to regular adjustment, for example the expense on the organization of work flow. Consider a situation of high monetary stability, followed by a period of lower monetary stability. Since the value of flexibility is lower in a stable environment, firms pay the setup cost and adopt regular adjustment. When the economy becomes less stable firms may not abandon regular adjustment since the setup cost has been paid. Therefore, even though the increased monetary stability is temporary, it permanently reduces flexibility of pricing policies at the firm level. Since monetary policy is more effective when firm follows regular adjustment, the result is a history-dependent slope of the Phillips curve.

References

- Álvarez, Luis and Ignacio Hernando, (2004): “Price Setting Behaviour in Spain. Stylised Facts Using Consumer Price Micro Data”, *European Central Bank Working Paper Series* No 416.
- Álvarez, Luis, Burriel, Pablo and Ignacio Hernando, (2005): “Do Decreasing Hazard Functions Make Sense?”, *European Central Bank Working Paper Series* No 461.
- Aucremanne, Luc and Emmanuel Dhyne (2004), “How Frequently Do Prices Change? Evidence Based on the Micro Data Underlying the Belgian CPI”, *European Central Bank Working Paper Series* No. 331.
- Aucremanne, Luc and Emmanuel Dhyne (2005), “Time-dependent versus State-dependent Pricing: A Panel Data Approach to the Determinants of Belgian Consumer Price Changes.”, *European Central Bank Working Paper Series* No 462.
- Basu, Kaushik (1997), “Why are so many goods priced to end in nine? And why this practice hurts the producers”, *Economics Letters* 54, 41-44.
- Baudry, Laurent, Hervé Le Bihan, Patrick Sevestre and Sylvie Tarrieu (2004): “Price Rigidity in France – Evidence from Consumer Price Micro-Data”, *European Central Bank Working Papers Series* No 384.
- Baumgartner, Josef, Ernst Glatzer, Fabio Rumler and Alfred Stiglbauer (2005): “How Frequently Do Consumer Prices Change in Austria? Evidence from Micro CPI Data”, *European Central Bank Working Paper Series* No 523.
- Bank of Canada (2000), “A New System of Fixed Dates for Announcing Changes to the Bank Rate”, <http://www.bankofcanada.ca/en/fixed-dates/>
- Bénabou, Roland (1992), “Inflation and Efficiency in Search Markets”, *The Review of Economic Studies*, 299-329.
- Bergen, Mark, Haipeng (Allan) Chen, Robert Kauffman, Dongwon Lee and Daniel Levy (2003), “Making Sense of Ignoring Cents: Price Points and Price Rigidity under Rational Inattention”, Bar-Ilan University and Emory University Working Paper, mimeo.
- Bils, Mark and Peter J. Klenow (2004), “Some Evidence on the Importance of Sticky Prices”, *Journal of Political Economy*, 947-85.
- Caplin, Andrew S. and Daniel F. Spulber (1987), “Menu Costs and the Neutrality of Money”, *Quarterly Journal of Economics*, 703-26.
- Caplin, Andrew and John Leahy (1991), “State-Dependent Pricing and the Dynamics of Money and Output”, *Quarterly Journal of Economics*, 683-708.
- Cecchetti, Stephen G. (1986), “The Frequency of Price Adjustment: A Study of the Newsstand Prices of Magazines”, *Journal of Econometrics*, 255-74.

- Dhyne, Emmanuel, Luis J. Álvarez, Hervé Le Bihan, Giovanni Veronese, Daniel Dias, Johannes Hoffmann, Nicole Jonker, Patrick Lünnemann, Fabio Rumler, Jouko Vilmunen (2005): “Price setting in the euro area: Some stylized facts from Individual Consumer Price Data”, *European Central Bank Working Paper Series* No 524.
- Dias, Daniel, Carlos Robalo Marques and Joao M. C. Santos Silva (2005): “Time or state dependent price setting rules? Evidence from Portuguese micro data”, *European Central Bank Working Papers Series* No 511.
- Dotsey, Michael, Robert G. King and Alexander L. Wolman (1999), “State-Dependent Pricing and The General Equilibrium Dynamics of Money and Output”, *Quarterly Journal of Economics* 114(2), 655-690.
- Dutta, Shantanu, Mark Bergen, Daniel Levy and Robert Venable (1999), “Menu Costs, Posted Prices, and Multi-product Retailers”, *Journal of Money, Credit, and Banking*, November, 1999, 683-703.
- Fischer, Stanley (1977), “Long-Term Contracts, Rational Expectations and the Optimal Money Supply Rule”, *Journal of Political Economy*, 191-205.
- Fisher, Timothy and Jerzy Konieczny. (2000), “Synchronization of Price Changes by Multiproduct Firms: Evidence from Canadian Newspaper Prices”, *Economics Letters*, 68, 271-277.
- Gagnon, Etienne (2006), “Price Setting under Low and High Inflation: Evidence from Mexico”, Northwestern University, mimeo.
- Hoffmann, Johannes and Jeong-Ryeol Kurz-Kim (2005): “Consumer price adjustment under the microscope: Germany in a period of low inflation”, Deutsche Bundesbank, mimeo.
- Jonker, N., C. Folkertsma and H. Blijenberg (2004), “An Empirical Analysis of Price Setting Behaviour in the Netherlands in the Period 1998-2003 Using Micro Data”, *European Central Bank Working Paper Series* No. 413.
- Konieczny, Jerzy D. and Andrzej Skrzypacz (2005), “Inflation and Price Setting: Evidence from a Natural Experiment”, *Journal of Monetary Economics*, 621-32.
- Konieczny, Jerzy D. and Andrzej Skrzypacz (2006), “Search. Costly Price Adjustment and the Frequency of Price Changes: Theory and Evidence”, mimeo.
- Levy, Daniel, Mark Bergen, Shantanu Dutta and Robert Venable (1997), “The Magnitude of Menu Costs: Direct Evidence from Large U.S. Supermarket Chains”, *Quarterly Journal of Economics*, 792-825.
- Levy, Daniel, Shantanu Dutta and Mark Bergen (2002), “Heterogeneity in Price Rigidity: Evidence from a Case Study Using Micro-Level Data”, *Journal of Money, Credit, and Banking* 34(1), 197–220.

- Levy, Daniel, Georg Müller, Shantanu Dutta and Mark Bergen (2005), "Holiday Price Rigidity and Cost of Price Adjustment", Bar-Ilan University and Emory University Working Paper, mimeo.
- Lünnemann, Patrick and Thomas Y. Mathä (2005): "Consumer price behaviour in Luxembourg: evidence from micro CPI data", *European Central Bank Working Papers Series* No 541.
- Niven, Ivan (1961), *Numbers: Rational and Irrational*, The Mathematical Association of America New Mathematical Library, Washington, D.C.
- Owen, Anne and David Trzepacz (2002), "Menu Costs, Firm Strategy, and Price Rigidity", *Economics Letters*, 76, 345-349.
- Papke, Leslie E. and Jeffrey M. Wooldridge (1996) "Econometric Methods for Fractional Response with an Application to 401(K) Plan Participation Rates", *Journal of Applied Econometrics*, Vol. 11, N.6, p. 619-632.
- Rotemberg, Julio J. (2005), "Customer Anger at Price Increases, Time Variation in the Frequency of Price Changes and Monetary Policy", *Journal of Monetary Economics*, 52, 829-52.
- Rotemberg, Julio J. (2006), "Fair Pricing", Harvard Business School, mimeo.
- Scarf, Herbert (1959), "The Optimality of (S, s) Policies in the Dynamic Inventory Problem", in: K. Arrow, P. Suppes and S. Karlin, Editors, *Mathematical Methods in the Social Sciences*, Proceedings of the First Stanford Symposium Stanford University Press, Stanford.
- Sheshinski, Eytan and Yoram Weiss (1977), "Inflation and Costs of Price Adjustment", *Review of Economic Studies*, 287-303.
- Taylor, John (1980), "Aggregate Dynamics and Staggered Contracts", *Journal of Political Economy*, 1-23
- Veronese, Giovanni, Silvia Fabiani, Angela Gattulli and Roberto Sabbatini (2005), "Consumer Price Behaviour In Italy: Evidence From Micro CPI Data", *European Central Bank Working Paper Series* No 449.
- Wolman, Alex (2000), "The Frequency and Costs of Individual Price Adjustment", Federal Reserve Bank of Richmond Economic Quarterly 86, 1-22.

Appendix

Definition of attractive prices

Attractive prices are defined for price ranges in order to take account of different attractive prices at different price levels: from 0 to 10 Austrian Schillings (ATS) all prices ending at x.00, x.50 and x.90 ATS, from 10 to 100 ATS all prices ending at xx0.00, xx5.00 and xx.90 ATS, from 100 to 1,000 ATS prices ending at xx0.00, xx5.00 and xx9.00 and xxx.90 ATS and so on. An equivalent rule has been defined to identify attractive prices in Euro after the cash changeover (2002-2003). Table A1 shows an excerpt of a list of attractive prices for the Schilling case. In order to give a complete list of attractive prices, the table would continue to the right and to the bottom. The extension to the right would show multiples of 10 and 100 of the last four columns.

Table A1: Attractive prices for the Schilling period (1996-2001)

below 1	1-9.99	10-99.99		100-999.99			1000-9999.99			
					105.00		1050.00			
0.50	1.00	10.00	10.90	100.00	109.00	100.90	1000.00	1090.00	1009.00	1009.90
					115.00	101.90		1150.00	1019.00	1019.90
0.90			11.90	110.00	119.00	102.90	1100.00	1190.00	1029.00	1029.90
					125.00	103.90		1250.00	1039.00	1039.90
			12.90	120.00	129.00	104.90	1200.00	1290.00	1049.00	1049.90
					135.00	105.90		1350.00	1059.00	1059.90
			13.90	130.00	139.00	106.90	1300.00	1390.00	1069.00	1069.90
					145.00	107.90		1450.00	1079.00	1079.90
			14.90	140.00	149.00	108.90	1400.00	1490.00	1089.00	1089.90
					155.00	109.90		1550.00	1099.00	1099.90
	1.50	15.00	15.90	150.00	159.00	110.90	1500.00	1590.00	1109.00	1109.90
					165.00	111.90		1650.00	1119.00	1119.90
			16.90	160.00	169.00	112.90	1600.00	1690.00	1129.00	1129.90
					175.00	113.90		1750.00	1139.00	1139.90
			17.90	170.00	179.00	114.90	1700.00	1790.00	1149.00	1149.90
					185.00	115.90		1850.00	1159.00	1159.90
			18.90	180.00	189.00	116.90	1800.00	1890.00	1169.00	1169.90
					195.00	117.90		1950.00	1179.00	1179.90
	1.90	19.00	19.90	190.00	199.00	118.90	1900.00	1990.00	1189.00	1189.90
					205.00	119.90		2050.00	1199.00	1199.90
	2.00	20.00	20.90	200.00	209.00	120.90	2000.00	2090.00	1209.00	1209.90
					215.00	121.90		2150.00	1219.00	1219.90
			21.90	210.00	219.00	122.90	2100.00	2190.00	1229.00	1229.90
					225.00	123.90		2250.00	1239.00	1239.90
			22.90	220.00	229.00	124.90	2200.00	2290.00	1249.00	1249.90
					235.00	125.90		2350.00	1259.00	1259.90
			23.90	230.00	239.00	126.90	2300.00	2390.00	1269.00	1269.90
					245.00	127.90		2450.00	1279.00	1279.90
			24.90	240.00	249.00	128.90	2400.00	2490.00	1289.00	1289.90
					255.00	129.90		2550.00	1299.00	1299.90
	2.50	25.00	25.90	250.00	259.00	130.90	2500.00	2590.00	1309.00	1309.90