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The Dynamics of European Inflation Expectations*

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Abstract

We investigate the relevance of Carroll's sticky information model of inflation expectations for four major European economies (France, Germany, Italy and the United Kingdom). In contrast to the most rational expectation models, households in the sticky information environment update their expectations occasionally rather than instantaneously due to the costs of acquiring and processing information. Using survey data on household and expert inflation expectations we argue that the model adequately captures the dynamics of household inflation expectations. We estimate two alternative parametrizations of the sticky information model which differ in the stationarity assumptions about the underlying series. Our baseline stationary estimation suggests that the average frequency of information updating for European households is roughly once in 18 months. The vector error-correction model implies households update information about once a year.

KEYWORDS: inflation expectations, sticky information, inflation persistence

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1 Introduction

Several recent papers (including Mankiw and Reis, 2002, 2006) argue that sticky information models, in which agents update their information occasionally rather than instantaneously, resolve some puzzles in the output–inflation dynamics. For example, sticky information models are able to account for considerable inflation persistence and substantial sacrifice ratios (recessionary disinflations) typically observed in the data.

Microeconomic foundations for the sticky information paradigm were elaborated in Carroll’s (2003) work on the “epidemiological model of expectations.” Carroll argues that US survey data on inflation expectations are consistent with a model in which in each period only a fraction of households adopts inflation forecasts of experts, whose forecasts are superior. The remaining households find it costly to update their information and continue using their own past expectations rather than form better predictions, an activity that requires more time and knowledge. In related work Sims (2003), Branch (2004), Reis (2006a), Reis (2006b) and others provide alternative justifications for models with agents that do not instantaneously incorporate all available information as implied by most standard macro models.¹

While the sticky information approach seems to be useful for modelling US data, corresponding evidence for European countries is, to the best of our knowledge, still lacking.² We attempt to fill this gap by investigating inflation expectation data from four major European economies (France, Germany, Italy and the United Kingdom).

Our findings in general support the usefulness of Carroll’s sticky information model for the description of expectational dynamics in Europe. We find that household inflation expectations adjust sluggishly to the more precise predictions of professional forecasters. The average speed of this adjustment varies little across the four countries in our sample and is somewhat lower than that in the US: a typical household updates its inflation expectations roughly once in eighteen months (compared to once a year previously found in the US). While this result is quite robust across the two estimation methods,

¹Other work on sticky information and price setting in the US includes Coibion (2007), Dupor et al. (2007), Kiley (2007), Klenow and Willis (2007), Laforte (2007) and Korenok (forthcoming).

²The only work testing sticky information models on international data is Khan and Zhu (2002), Handjiyska (2004) and Korenok et al. (2006). However, these papers have to adopt some restrictive assumptions to circumvent data limitations: Khan and Zhu approximate agents’ expectations with forecasts from a VAR model. Handjiyska uses interpolated data on expert expectations. Korenok et al. use actual data from 13 OECD economies (not data on expectations).

we find that the frequency of information updating in Europe is somewhat higher for the vector error-correction specification, amounting to about once a year. Similarly to the US, European households are not backward-looking: they tend to update their expectations from experts' rational forecasts rather than actual past inflation rates. Finally, the estimates are stable over time: we cannot distinguish any statistically significant differences between various institutional settings (e.g., inflation-targeters and non-targeters).

For policy-makers these results imply that inflation expectations of (European) consumers are sluggish. Consequently, the channel from household expectations to actual inflation rates is likely to remain an important source of inflation persistence even if expert expectations are rational in that they adjust instantaneously to new information. Higher central bank credibility has in the sticky information setup an ambiguous effect on the persistence of household expectations: while expert expectations become more tightly anchored to inflation target, households update their expectations less often. The first effect thus decreases the persistence of household inflation expectations; the second works in the opposite direction.

The remainder of the paper is structured as follows. Section 2 briefly discusses the theoretical model of household expectations proposed by Carroll (2003). Section 3 introduces the data sets we use in the empirical part. Section 4 elaborates on econometric issues and presents the empirical results of our analysis. Section 5 concludes.

2 The Epidemiology of Household Inflation Expectations

Carroll (2003) proposes the following micro-founded model of the transmission of inflation expectations. The economy consists of two types of agents: experts (professional forecasters) and households. Experts collect in every period relevant information on future inflation, make superior inflation forecasts and publish them in newspapers. Because reading newspapers (or making informed inflation forecasts) is costly, households—in contrast to the standard frictionless rational framework—choose to update their expectations occasionally rather than instantaneously. As a result, new information about inflation spreads slowly across population in the following “epidemiological” way. In each period only a randomly chosen fraction λ_1 of households follows the latest inflation stories and updates its inflation expectations. The remaining $1 - \lambda_1$ households stick to their forecasts from the previous period. The evolution of

the (average) household (denoted HH) inflation (π) expectation (\mathbf{E}) follows:

$$\mathbf{E}_t^{HH} \pi_{t,t+1} = \lambda_1 \mathbf{E}_t^{EX} \pi_{t,t+1} + (1 - \lambda_1) \mathbf{E}_{t-1}^{HH} \pi_{t,t+1},$$

where $\mathbf{E}_t^{HH} \pi_{t,t+1}$ and $\mathbf{E}_t^{EX} \pi_{t,t+1}$ denote one-period-ahead inflation expectations of households and experts (EX), respectively.

Thus, news about inflation can be thought of as a disease that spreads slowly across the population, infecting a fraction λ_1 of all households in each period. The calculation outlined in Carroll (2003) leads to a similar equation formulated for expected one-year-ahead inflation rates collected at quarterly frequency, which are typically reported in surveys of inflation expectations:

$$\mathbf{E}_t^{HH} \pi_{t,t+4} = \lambda_1 \mathbf{E}_t^{EX} \pi_{t,t+4} + (1 - \lambda_1) \mathbf{E}_{t-1}^{HH} \pi_{t-1,t+3}. \quad (1)$$

Equation (1) holds if (i) inflation follows a random walk process (or households believe that inflation is a random walk) or (ii) $\mathbf{E}_{t-1}^{HH} \pi_{t-1,t+3} \approx \mathbf{E}_{t-1}^{HH} \pi_{t,t+4}$. Both of these assumptions are likely to be satisfied in our dataset. As discussed below, the underlying CPI inflation process in the core European economies has, indeed, been very persistent recently, warranting the random walk approximation. In addition, given the high persistence of the inflation process, there is not much difference between household expectations as of time $t - 1$ of inflation rates at $t + 3$ and $t + 4$, which, in turn, implies that condition (ii) is also likely to be met.

3 Data on Inflation Expectations

To test the model of information diffusion, we need two inflation expectations series: inflation forecasts of households and professional forecasters. The forecasts of households were obtained from the European Commission's (EC) consumer survey and those of professional forecasters from Consensus Economics, a London-based macroeconomic survey firm.³

Household expectations were constructed using the EC survey's question 6, which asks how, by comparison with the last 12 months, the respondents expect that consumer prices will develop in the next 12 months.⁴ Unfortun-

³We only investigate Germany, France, Italy and the United Kingdom because expectations of professional forecasters from other European countries, such as the Netherlands and Spain, are not available before 1996.

⁴The exact wording of question 6 of the Consumer Survey of the Joint Harmonised EU Programme of Business and Consumer Surveys is: "By comparison with the past 12 months, how do you expect that consumer prices will develop in the next 12 months?" For more information on the survey, see the Commission's web page, http://ec.europa.eu/economy_finance/db_indicators/db_indicators8650_en.htm.

nately, the answers are qualitative rather than quantitative (unlike, for example, the question on expected inflation in the US Michigan Survey of Consumer Sentiment). This means that the respondents are asked about the direction of the expected movement of consumer prices (increase/fall), not about the exact value of this movement. Consequently, care needs to be taken when transforming these data into quantitative measures of household expectations, required to test equation (1). We follow much of the existing literature (including Gerberding, 2001 and Mankiw et al., 2003) in adopting the Carlson and Parkin (1975) method, explained in the Appendix.

To appropriately rescale the expectations it is necessary to specify a variable that captures the perceived current level of the inflation rate. We investigated three alternatives that have been proposed in the literature: (i) recursive Hodrick–Prescott filter, (ii) past inflation and (iii) inflation extracted as fitted values from the regression of inflation on the past balance statistics from the survey (lagged by four quarters). The normalization that works best in terms of (low) mean squared error and is used here is the recursive Hodrick–Prescott filter, in which the inflation trend was extracted in a quasi-real-time way.⁵

Figure 1 compares expert and household inflation expectations with actual inflation rates. Most of the time both expert and household predictions are close to actual inflation. However, sometimes there are rather persistent differences between expectations and actual inflation. More importantly, because household and expert expectations occasionally differ quite considerably (such as in the early 1990s in the UK), a closer examination of the dynamic interaction of both variables is warranted.

4 Empirical Results

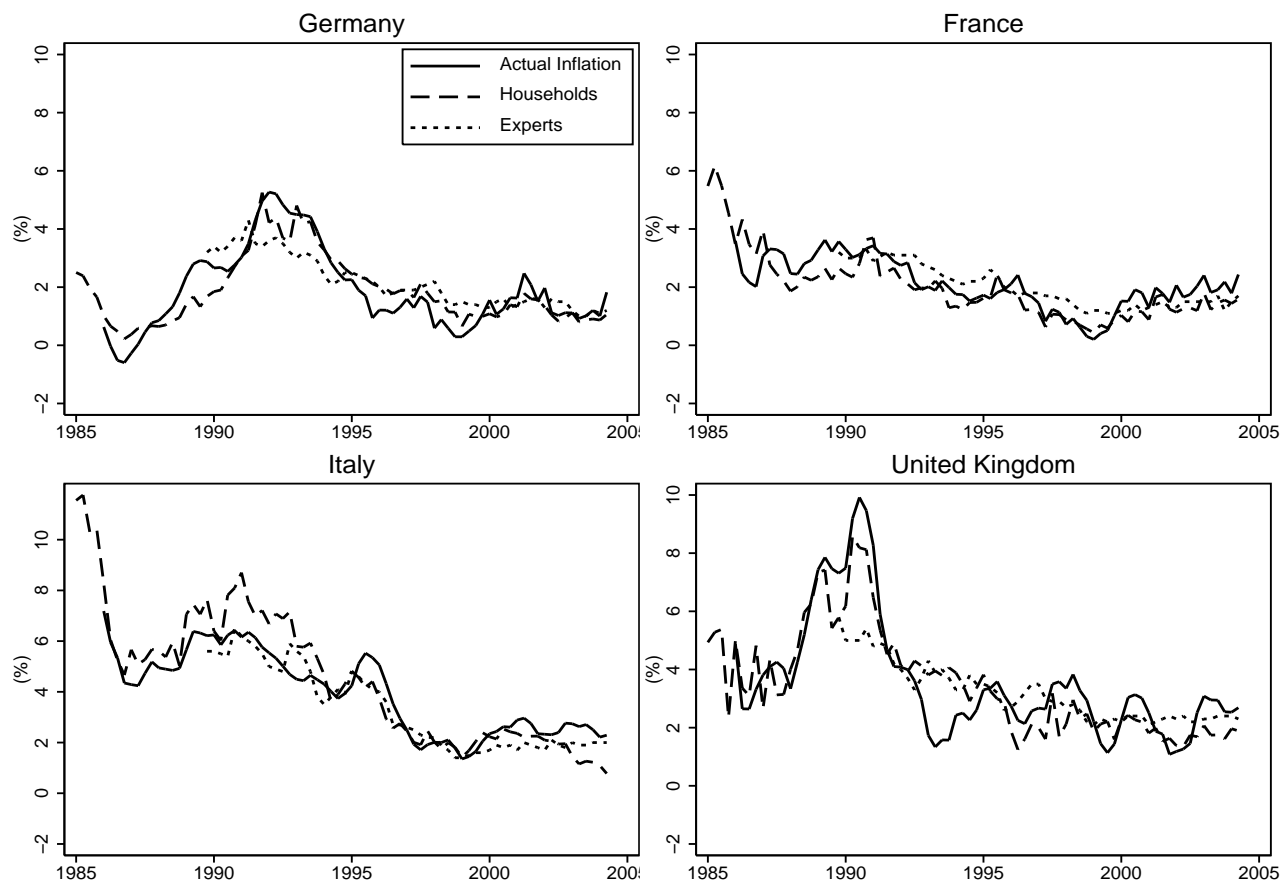
The choice of the appropriate empirical strategy depends on the time series properties of the underlying expectations. If the series are stationary, equation (1) can be estimated directly using OLS (as in Carroll, 2003). If they are non-stationary (I(1)) and cointegrated, the model should be transformed into the vector error-correction (VEC) form.

A number of recent papers investigate the degree of persistence of various measures of inflation in Europe.⁶ Although these studies agree that inflation is

⁵The details, beyond those outlined in the Appendix, are available from the authors on request. The results reported in the paper typically hold for alternative normalizations considered. In particular, we find that the alternative rescaling procedures typically imply that λ_1 is positive and significant and the summing-up restriction $\lambda_1 + \lambda_2 = 1$ is met (using the same notation as, e.g., equation (2) below).

⁶See European Central Bank (2005), table 3.4, page 21 for a summary of the literature

Figure 1: Household and Expert Expectations and Actual Inflation



a very persistent process, the evidence on its order of integration (i.e., whether it is stationary or $I(1)$) is less conclusive. Most papers cited in European Central Bank (2005) reject the null hypothesis that inflation in large European countries has recently been non-stationary. In contrast, the recent work of O'Reilly and Whelan (2005) (on inflation in the euro area) as well our preliminary tests (investigating inflation and inflation expectations in the EU-4 countries) in general do not reject the null.⁷ A potential criticism of our results is that the sample is too short to allow reliable inference. The fact that we are unable to reject the null may well result from the notoriously low power of the unit root tests under such circumstances, rather than the existence of the unit root.

Since the main focus of this paper is not on providing the definitive answer on the order of integration of inflation (or inflation expectations), we now move on to estimating Carroll's theoretical model and investigating how sensitive its implications are depending on whether we assume stationary or non-stationary environments. Because the tests do not clearly determine the stationarity properties in the relatively short sample we have, we first estimate the Carroll (2003) model in the stationary framework. We then briefly consider how the results are affected if the nonstationary (VEC) setup is adopted.

4.1 The Stationary Case: The Carroll Model

Before estimating equation (1), we will examine some preliminary evidence on the relationship between expert and household expectations. Given the interest in the interaction between the expectations of both professional forecasters and households, a natural starting point is to ask, (i) which of the two group forecasts is (on average) better and (ii) what the Granger causality is between the two expectations. We provide the answers in the next subsection.

Relationship Between Expert and Household Expectations

First, expert expectations are substantially more precise than household expectations. The root mean squared errors of expert forecasts are between 15% to 35% lower in Germany, Italy and the UK than for household expectations. The two expectations are comparably precise in France.⁸ This result does

on European countries.

⁷The results are available from the authors on request.

⁸This is in line with the findings of Gerberding (2001), who investigates rationality, efficiency and unbiasedness of household and expert inflation expectations in detail. She reports that while expert expectations have lower RMSEs (except for France), neither household nor

Table 1: Tests for Granger Non-causality

$$\mathbf{E}_t^i \pi_{t,t+4} = \beta_0 + \sum_{j=1}^p \beta_j \mathbf{E}_{t-j}^{EX} \pi_{t-j,t+4-j} + \sum_{k=1}^p \gamma_k \mathbf{E}_{t-k}^{HH} \pi_{t-k,t+4-k} + \varepsilon_t$$

Country	Dependent Variable: Expectations of . . .	$\beta_j = 0, \forall j$ p-value	$\gamma_k = 0, \forall k$ p-value
Germany	Experts	0.000	0.125
	Households	0.000	0.000
France	Experts	0.000	0.076
	Households	0.000	0.000
Italy	Experts	0.000	0.010
	Households	0.620	0.000
United Kingdom	Experts	0.000	0.149
	Households	0.009	0.000

Notes: Sample: 1989:4 to 2004:2. The tests were computed with $p = 2$ lags of independent variables.

not, of course, come as a surprise since households may know expert forecasts when forming their own expectations. According to the epidemiology model, at least some households update their own expectations by following experts. In addition, in an environment with stable inflation like that in France in the post-1980s, neither forecast varies much.

Second, we can examine in table 1 whether expert forecasts Granger-cause household forecasts by testing for joint significance of the appropriate coefficients in the following equations:

$$\mathbf{E}_t^i \pi_{t,t+4} = \beta_0 + \sum_{j=1}^p \beta_j \mathbf{E}_{t-j}^{EX} \pi_{t-j,t+4-j} + \sum_{k=1}^p \gamma_k \mathbf{E}_{t-k}^{HH} \pi_{t-k,t+4-k} + \varepsilon_t,$$

where the regressions are alternatively estimated with both expert and household expectations as dependent variables, $i \in \{EX, HH\}$. Columns 3 and 4 indicate that lags of expert expectations are typically significant predictors of household expectations. Household expectations, on the other hand, tend not to Granger-cause experts. Thus, in all countries, except for Italy we conclude that the direction of Granger causality goes from experts toward households.⁹

expert expectations are fully efficient (and can be improved upon with available information on past inflation, unemployment and interest rates).

⁹Admittedly, the p-values on lagged household expectations in expert equations in the upper right corner of each country-specific panel, which range between 0.08 and 0.15, are

Table 2: Sticky Expectations in Europe I.
Restricted Cross-Country Results

$$\mathbf{E}_t^{HH} \pi_{t,t+4} = \lambda_0 + \lambda_1 \mathbf{E}_t^{EX} \pi_{t,t+4} + \lambda_2 \mathbf{E}_{t-1}^{HH} \pi_{t-1,t+3} + \lambda_3 \pi_{t-5,t-1} + \varepsilon_t$$

Model	λ_0	λ_1	λ_2	λ_3	Test	Cross eqn
					p-value	p-value
M1		0.17*** (0.04)	0.83*** (0.03)		$\lambda_1 + \lambda_2 = 1$ 0.912	0.04
M2		0.17*** (0.03)	0.83*** (0.03)		$\lambda_1 = 0.25$ 0.016	0.62
M3	-0.22*** (0.07)	0.29*** (0.05)	0.78*** (0.03)		$\lambda_0 = 0$ 0.003	0.15
M4		0.31*** (0.05)	0.77*** (0.05)	0.00 (0.04)	$\sum_{i=1}^3 \lambda_i = 1$ 0.003	0.03
M5	-0.22*** (0.07)	0.29*** (0.05)	0.78*** (0.05)	-0.01 (0.04)	$\lambda_3 = 0$ 0.900	0.13
M6			0.92*** (0.04)	0.05 (0.04)	$\lambda_2 + \lambda_3 = 1$ 0.015	0.34

Notes: Sample: 1989:4 to 2004:2. Seemingly unrelated regressions. Standard errors in brackets. *, **, *** denotes rejection of the null at the 10%, 5% and 1% significance level, respectively.

What Determines Household Expectations?

Having found supportive preliminary evidence for the epidemiological model of expectation formation, let us now turn to direct estimation of and inference about the speed of information updating, λ_1 . Table 2 summarizes the estimation results of various constrained versions of the following regression:

$$\mathbf{E}_t^{HH} \pi_{t,t+4} = \lambda_0 + \lambda_1 \mathbf{E}_t^{EX} \pi_{t,t+4} + \lambda_2 \mathbf{E}_{t-1}^{HH} \pi_{t-1,t+3} + \lambda_3 \pi_{t-5,t-1} + \varepsilon_t. \quad (2)$$

All models are estimated with seemingly unrelated regressions (SUR) with coefficients restricted to be equal across the four countries.¹⁰

quite low, which may suggest the existence of some feedback from households to experts. An additional piece of evidence supporting the causality from experts to households comes from the VEC setup of equation (4) below: While the loading coefficients in the household equations are, as reported in table 5 below, significant, in the expert equations they are statistically indistinguishable from zero or have the wrong (positive) sign.

¹⁰Analogous results hold when the models are estimated with equation-by-equation OLS. However, since the cross-correlation between residuals in our dataset is up to 0.3, SUR improve efficiency of our estimates. In addition, SUR also make it possible to test (and impose)

The format of table 2 follows that of Carroll (2003), table III. The left panel (the first four columns) displays the point estimates of λ_1 s together with standard errors; the right panel shows specification tests (p-values of various tests of coefficients). The last column tests whether the coefficients are the same across countries.¹¹

The models are labelled M1–M6. The first model, M1, estimates the following version of (2):

$$\mathbf{E}_t^{HH} \pi_{t,t+4} = \lambda_1 \mathbf{E}_t^{EX} \pi_{t,t+4} + \lambda_2 \mathbf{E}_{t-1}^{HH} \pi_{t-1,t+3} + \varepsilon_t, \quad (3)$$

in which coefficients λ_1 and λ_2 are estimated as unrestricted. The estimates of λ_1 and λ_2 are 0.17 and 0.83, respectively; both are statistically significant. The summing-up restriction implied by the Carroll model, $\lambda_1 + \lambda_2 = 1$, is clearly satisfied.

Model M2 is estimated for the restricted version with the summing-up restriction imposed. Given how close the restriction is to being met in M1, it does not come as a surprise that the point estimates of λ_1 barely change.¹²

Our baseline $\lambda_1 = 0.17$ suggests that an average European household reads economic updates or consults economic experts roughly once in 18 months.¹³ In addition, the estimate implies about 47% of households use information outdated more than one year and about 23% more than two years. Thus household expectations seem to react very sluggishly to the available macroeconomic news (that has already been processed by forecasters). In fact, our

cross-equation restrictions and answer questions such as: “Does the speed of information updating vary across countries?”

We report some results unrestricted across countries below in table 3. More detailed results of unrestricted SUR and equation-by-equation estimation are available from the authors on request. The results are robust to these modifications.

¹¹To conserve space we do not report measures of fit, which of course differ slightly for each country (and model). \bar{R}^2 s vary between 0.75 and 0.96. For more information of how well selected models explain household inflation expectations see table 3 below.

Similarly to Carroll’s findings, the Durbin–Watson statistics detect virtually no autocorrelation in residuals for models M1–M6.

It is well-known that OLS estimates in a setup with lagged dependent variables and autocorrelated residuals are inconsistent. To address this potential concern we have reestimated the models allowing for first-order autocorrelation in disturbances and found no statistically significant evidence (at the 5 % level) of autocorrelated errors.

¹²That the summing up restriction holds is not particularly surprising because it also holds outside the epidemiology framework. For example, this restriction is met in any model which implies that the household and expert expectations are cointegrated with a cointegrating vector $(1, -1)^\top$.

¹³This frequency is calculated as $1/\lambda_1 \approx 6$ quarters.

estimate λ_1 is very much lower than one, which suggests the costs of acquiring new macroeconomic information are substantial (relative to the benefits).¹⁴

The speed of adjustment $\lambda_1 = 0.17$ is lower than Carroll's baseline coefficient of 0.27. Because the standard error on λ_1 is small, the difference is statistically significant. However, much of the difference between λ_1 in Europe and the US can be accounted for with different time ranges: Carroll's sample (1981:3–2000:2) differs from ours (1989:4–2004:2). Re-estimating model M2 with the US data and our sample range gives $\lambda_1 = 0.16$ (for the US). This matches Carroll's evidence that updating is faster when inflation is in the news, including the early period in his sample. In contrast, in the 1990s λ_1 has fallen, because inflation (in the US) received less press coverage than before 1985. In addition, the recent monetary policy leading to low and stable inflation has reduced uncertainty and, together with smaller exogenous shocks hitting the economy, has presumably further decreased households' incentive to update.

Models M3–M6 investigate four alternative structures of household expectations. First, we add a constant to equation (3). This parameter turns out to be negative and significantly different from zero. As we will see below (in table 3), this result, which contradicts the simple epidemiology model (1), is driven mainly by the early sample in the UK where (around 1991) the actual inflation rate was falling considerably. In such an environment the epidemiology model (1) may not be adequate in that the non-updating $(1 - \lambda_1)$ households decide to adjust for the falling trend in inflation by adding a negative number to their past forecasts $\mathbf{E}_{t-1}^{HH} \pi_{t-1,t+3}$, which may then cause the negative estimate of λ_0 in table 2 (and for the UK in table 3 below).¹⁵

Models M4–M6 allow for the possibility that consumers are at least in part backward-looking (adaptive) by adding past inflation to the right-hand side of (2). Similarly to the US, there is very little of the backward-looking element in household inflation expectations: the coefficient λ_3 is small both in terms of its size and its significance level. Thus, households recognize that when reading newspapers it makes more sense to learn from the forward-looking experts than to naively extrapolate the past inflation rates.

¹⁴The finding that uncertainty about aggregate variables matters little compared with idiosyncratic uncertainty is in line with the independent evidence on the small cost of eliminating business cycles, reported, e.g., in Krusell and Smith (1999), and in the sticky information setup with consumption dynamics by Carroll and Slacalek (2006).

¹⁵One piece of evidence that is consistent with this hypothesis is that reestimating model M3 for the post-1992 sample gives an insignificant $\lambda_0 = -0.05$ (with the standard error of 0.07). (Estimating the same “unrestricted” model of table 3 below for the same sample also gives for the UK an insignificant constant $\lambda_0 = -0.14$ with the standard error of 0.31.)

Other, more general specifications of the epidemiology model could be considered. For example, how does augmenting the model (3) with more lags of expert and/or household expectations affect the estimates of λ_1 ? The results not reported in the paper suggest that adding an additional lag of household expectations does not affect the results much (because the coefficient is small and statistically indistinguishable from zero). Adding a lag of expert expectations ($\mathbf{E}_t^{EX} \pi_{t-1,t+3}$) causes the standard error on λ_1 to increase considerably and coefficients on expert expectations to become individually insignificant; jointly, however, they remain very significantly positive (and sum close to the number implied by the epidemiology model).

Generally, there appears to be a lot of homogeneity across countries. As indicated in the last column, in four of the six (M1–M6) models considered the null of equal coefficients in the four countries is satisfied; two models (M1 and M4) yield borderline rejections of homogeneity (at the 5% significance level).

Cross-Country Differences

Having found supportive evidence for Carroll’s sticky information model in European data, let us now investigate in more detail how the findings vary across countries. Table 3 summarizes estimation results obtained from seemingly unrelated regressions, unrestricted across countries, for models M1–M3.

The findings parallel those in the previous section: First, the speed of updating λ_1 varies between 0.11 and 0.32 (as estimated with models M1 and M2) and is for all countries except Italy highly statistically significant. Second, for all countries, except France, the summing-up restriction, $\lambda_1 + \lambda_2 = 1$, implied by Carroll’s sticky information model, is met. Even for France, the two coefficients sum up to 0.91. Third, the intercept term λ_0 is insignificant for all countries except the UK.

The estimated speed of updating is only about half as large in Italy as in the other countries. This finding seems counterintuitive because the benefits from acquiring information in an environment with a relatively high level and uncertainty about inflation should actually be larger. We believe the econometric result is driven by the persistent differences between household and professionals’ expectations at the beginning of the sample, before 1993 or so (apparent in figure 1). These deviations then may translate into a lower estimate λ_1 . To investigate this hypothesis, we have reestimated the regressions for the post-1992 period and found the Italian λ_1 substantially higher—about 0.25—and more significant.

It is a bit surprising that the speed of updating does not seem to vary more

Table 3: Sticky Expectations in Europe II.
Country-by-Country Unrestricted Results

$$\mathbf{E}_t^{HH} \pi_{t,t+4} = \lambda_0 + \lambda_1 \mathbf{E}_t^{EX} \pi_{t,t+4} + \lambda_2 \mathbf{E}_{t-1}^{HH} \pi_{t-1,t+3} + \varepsilon_t$$

Model	λ_0	λ_1	λ_2	\bar{R}^2	Test p-value
Germany					
M1		0.18*** (0.06)	0.82*** (0.06)	0.91	$\lambda_1 + \lambda_2 = 1$ 0.764
M2		0.20*** (0.06)	0.80*** (0.06)	0.91	$\lambda_1 = 0.25$ 0.368
M3	-0.21* (0.12)	0.29*** (0.08)	0.80*** (0.06)	0.91	
France					
M1		0.32*** (0.08)	0.59*** (0.09)	0.85	$\lambda_1 + \lambda_2 = 1$ 0.002
M2		0.18*** (0.07)	0.82*** (0.07)	0.83	$\lambda_1 = 0.25$ 0.322
M3	-0.04 (0.12)	0.33*** (0.10)	0.61*** (0.09)	0.80	
Italy					
M1		0.14 (0.11)	0.86*** (0.09)	0.96	$\lambda_1 + \lambda_2 = 1$ 0.991
M2		0.11* (0.06)	0.89*** (0.06)	0.96	$\lambda_1 = 0.25$ 0.022
M3	-0.18 (0.15)	0.25* (0.13)	0.81*** (0.09)	0.95	
United Kingdom					
M1		0.23*** (0.08)	0.77*** (0.08)	0.89	$\lambda_1 + \lambda_2 = 1$ 0.763
M2		0.23*** (0.08)	0.77*** (0.08)	0.89	$\lambda_1 = 0.25$ 0.781
M3	-0.67** (0.30)	0.53*** (0.16)	0.66*** (0.09)	0.89	

Notes: Sample: 1989:4 to 2004:2. Seemingly unrelated regressions. Standard errors in brackets. *, **, *** denotes rejection of the null at the 10%, 5% and 1% significance level, respectively.

across countries, which are in many respects very much unlike each other.¹⁶ Distinct differences may show up more clearly once more data are available.

We could now in principle similarly test how stable λ_1 has been over time. However, due to the limited number of observations the tests of structural stability have weak power. We looked for differences in the speed on expectation updating by including dummy variables for, e.g., members of the euro area and inflation-targeters. Unfortunately, these investigations did not lead to any statistically clear-cut results about the determinants of λ_1 .

Most of our findings suggest that the epidemiology model of information diffusion performs similarly well, qualitatively as well as quantitatively, for the core European countries as it does for the US. Expert inflation expectations are typically more precise than household expectations. Econometric tests indicate that the Carroll model is adequate along several dimensions.¹⁷ Most models imply that European households update somewhat more slowly than US households, on average once in 18 months compared with once a year. Finally, there is strong evidence that, as suggested by the epidemiology model, European households update information from the professional forecasters rather than the past inflation rate.¹⁸

¹⁶At least two kinds of institutional differences are potentially relevant for the size of λ_1 . First, the monetary policy setup and recent inflation dynamics in various EMU countries, the UK and the US are quite varied. For example, whereas Germany, under the Bundesbank regime, has always had moderate and stable inflation rates, Italy faced considerably higher inflation rates in the early 1990s and has witnessed pronounced declines in price level increases over the past decade in the run-up to and after the introduction of the euro. In addition, different communication strategies of central banks might affect how information spreads across households. (See for example Ehrmann and Fratzscher, 2005 and the literature cited therein.) Second, both the size and structure of the forecasting industry are dissimilar. This profession is in the US and the UK dominated by private forecasters, while in much of continental Europe, in particular in Germany, public forecasters play a more prominent role.

¹⁷For example, the speed of updating is positive and statistically significant, the summing-up restriction holds fairly well and household inflation expectations are not sensitive with respect to the past inflation.

¹⁸Consideration might be given to the possibility that households update their expectations by referring directly to other publicly available information, such as foreign prices. However, in the epidemiology framework this information is already captured and processed by professional forecasters, who are assumed to be rational. Moreover, obtaining such information is presumably much more costly than simply referring to the published professional forecasts.

Econometric Issues: Measurement Error and Endogeneity

One concern about our SUR estimates is that they are affected by measurement errors in household inflation expectations (caused because household expectations have to be rescaled). In presence of such classical measurement error coefficients λ_2 for model M1 in table 3 are biased toward zero. To get a feel for how serious this bias is we have reestimated the models of table 3 using past household and expert expectations as instruments for (the potentially mismeasured) $\mathbf{E}_t^{HH} \pi_{t-1,t+3}$. Compared to the SUR procedure the instrumental variables estimates of λ_2 fall roughly from 0.8 to 0.6. Together with this falling coefficient on lagged household expectations, the estimated speed of consumers' information updating λ_1 rises from about 0.2 to 0.4.

Another potentially relevant econometric issue with equations like (1) is endogeneity:¹⁹ Because experts presumably take into account households' expectations when making forecasts, they are affected by shocks to households' beliefs, which may induce correlation between $\mathbf{E}_t^{EX} \pi_{t,t+4}$ and the error term ε_t . We have attempted to address this issue by estimating the regressions instrumenting current expert expectations with their lags and found that the estimates were similar to those shown in table 3.

4.2 The Carroll Model in the Vector Error-Correction Form

Having estimated the epidemiology model in the stationary framework, let us now examine how the implications change when we assume that the expectation series are I(1) instead. Suppose we collect expert and household expectations in a vector $x_t = (\mathbf{E}_t^{HH} \pi_{t,t+4}, \mathbf{E}_t^{EX} \pi_{t,t+4})^\top$. If the two series are cointegrated with the cointegrating vector $\alpha = (1, -\alpha_1)^\top$, the system has the following vector error correction (VEC) representation:

$$\Delta x_t = \lambda_1 \alpha^\top x_{t-1} + \beta(L) \Delta x_t + \varepsilon_t, \quad (4)$$

where $\lambda_1 = (\lambda_1^{HH}, \lambda_1^{EX})^\top$ denotes the vector of loading coefficients and $\beta(L)$ is a matrix lag polynomial. Similarly to the stationary model (1), λ_1 determines the speed of adjustment toward the (long-run) equilibrium.²⁰ We are particularly interested in λ_1^{HH} , which corresponds to the speed of adjustment

¹⁹We thank a referee for bringing up this point.

²⁰The adjustment pattern in the partial adjustment version of the model (1), however, differs from the VEC analysis in two ways: First, the adjustment in the VEC is analyzed in an interdependent system and feedback effects are considered and second, the short-run dynamics in the VEC might influence the dynamic adjustment path.

Table 4: Tests for Cointegration between Household and Expert Expectations

Unrestricted	Cointegration	Rank	Test	(Trace)
Hypothesized No. of CE(s)	Eigenvalue	Trace Statistic	5% Critical Value	p-value [†]
Germany				
None	0.20	16.31	12.32	0.01
At most 1	0.06	3.48	4.13	0.07
France				
None	0.22	15.92	12.32	0.01
At most 1	0.03	1.96	4.13	0.19
Italy				
None	0.36	21.69	12.32	0.00
At most 1	0.09	3.65	4.13	0.07
United Kingdom				
None	0.25	18.90	12.32	0.00
At most 1	0.05	2.60	4.13	0.13

Note: [†] MacKinnon–Haug–Michelis (1999) p-values. Sample: 1989:4 to 2004:2, Italy: 1992:4–2002:4.

observed for households. Furthermore, note that the theoretical derivation of the epidemiology model predicts a cointegrating vector $\alpha = (1, -1)^\top$. This is due to the fact that in the long-run households completely adapt to the professional forecasts.

Before estimating the VEC representation (4) and its “ α -restricted” counterpart some preliminary specification tests need to be done. First, we test whether there exists a valid cointegrating relationship between the expert and household expectations as shown in table 4. The findings show that, for all four countries, the two series are cointegrated (at the 5% significance level). In addition, we checked whether the theoretical restriction on $\alpha = (1, -1)^\top$ is supported in data. The values for α_1 are close to -1 (the value predicted by the model) and range from -1.21 for the UK to -1.00 for Germany. As illustrated by the likelihood ratio statistics presented in table 5, we find that α is not significantly different from $(1, -1)^\top$ (except in the UK).

The VEC findings are summarized in table 5.²¹ All estimates of λ_1^{HH} are significant (although for the restricted case in France and Italy only at the 10% level) and lie except for the restricted case in Italy in the neighborhood

²¹The models were estimated for the time frame between 1989:4 and 2004:2, except for Italy, where a valid cointegrating relationship was found between 1992:4 and 2002:4.

Table 5: Sticky Expectations in the VEC Form

Model		Germany	France	Italy	UK
Unrestricted	$\hat{\lambda}_{HH}$	-0.30***	-0.26***	-0.18**	-0.20***
	std. error	(0.09)	(0.09)	(0.07)	(0.05)
Restricted	$\hat{\lambda}_{HH}$	-0.30***	-0.14*	-0.23*	-0.27***
	std. error	(0.09)	(0.08)	(0.12)	(0.08)
Test for restriction (1, -1) on α	LR stat.	0.00	2.29*	2.97*	3.86**
	p value	0.988	0.070	0.085	0.049

Notes: Sample: 1989:4 to 2004:2, Italy: 1992:4–2002:4. “Unrestricted” refers to the unrestricted VEC model. “Restricted” refers to the VEC estimation results under the restriction $\alpha = (1, -1)^\top$. *, **, *** denotes rejection of the null at the 10%, 5% and 1% significance level, respectively.

of 0.25, typically somewhat higher than implied by the “stationary” results above. We again find a lot of homogeneity among the four countries with French and Italian households updating presumably somewhat slower than British and German ones. The estimated updating frequencies in table 5 lie between once in three and seven quarters.

Carroll’s sticky information model is also supported by how the deviations from the long-term equilibrium are corrected.²² The error-correction process is primarily driven by the adjustment in household rather than expert expectations. This is implied by the estimates of λ_1^{EX} , which are not significantly different from zero or have the wrong (positive) sign.

Our findings thus imply that the epidemiology model of Carroll (2003) can be easily extended to the “non-stationary world.” The derived VEC epidemiology model of information diffusion performs similarly well to the stationary model. This result is especially useful for the analysis of European countries, which plausibly have highly persistent inflation rates (see O’Reilly and Whelan, 2005 and references in European Central Bank, 2005). Thus, even though it seems to be difficult to draw clear conclusions about the stationarity properties of the series with the small sample size at hand, the VEC representation might be preferable once more data are available.

²²Detailed results are available from the authors on request.

5 Conclusion

Inflation expectations are crucial determinants of future inflation dynamics. The model estimated here attempts to analyze how these expectations are formed and how information is transmitted from professional forecasters to households. Our estimates of the speed of information updating have important implications for the persistence of inflation and inflation expectations. We document that the qualitative and quantitative findings previously reported for the US generalize to major European countries. Most European households adjust rather sluggishly to new information; they update their information on average once in twelve to eighteen months. In addition, it turns out that households are forward-looking in that they use forecasts provided by experts rather than just past information to update their beliefs about future inflation. These findings are robust to the two estimation methods (suited for data with various stochastic properties) we consider.

The research in this paper can be extended through a number of avenues. Survey data can be used to directly estimate the sticky-information Phillips curve in addition to its epidemiological micro-foundations (as in Döpke et al., forthcoming). Alternatively, it would be possible, in the spirit of Mankiw et al. (2003), to analyze the evolution of the cross-sectional distribution of inflation expectations in Europe rather than just their mean values. Finally, the epidemiology model could, in principle, be estimated for additional countries, using cross-sectional dependence among countries to alleviate problems related to short samples.

Appendix—Data Issues

Expert Forecasts

The professional forecasts were obtained from Consensus Economics, a London-based macroeconomic survey firm. The survey of experts of private and public institutions in major industrial countries has been collected monthly since 1989. Once every quarter the questionnaire contains a question on forecasts over the next six quarters. The consensus forecast, used in the paper as a measure of expert expectations, is the mean of about 20 to 30 forecasts of local experts from major banks or research institutes in each country.

Household Forecasts

Our measures of household inflation forecasts are based on disaggregated answers to question 6 from European Commission's Harmonised Business and Consumer Surveys. The sample size of the survey is about 2,000 households in Germany, Italy and the UK, and roughly 3,300 households in France. The answers are reported as balance statistics and are available monthly since 1985.

Extracting Household Inflation Expectations

To obtain a measure of inflation expectations we have to re-scale the balance statistics. The standard method follows Carlson and Parkin (1975) and its extensions (see, for example, Gerberding, 2001, Mankiw et al., 2003 and Nielsen, 2003). The observed data are from a pentachotomous survey, i.e., they classify the responses into five subgroups:

Consumer prices will:

- Increase more rapidly,
- Increase at the same rate,
- Increase at a slower rate,
- Stay about the same,
- Fall.

Batchelor and Orr (1988) derive how responses from a pentachotomous survey can be transformed into a measure of inflation expectations: ${}_t\mu_{t+1} = \tilde{\mu}_t \times f({}_tA_{t+1}, \dots, {}_tE_{t+1})$, where ${}_tA_{t+1}, \dots, {}_tE_{t+1}$ are the fractions of respondents answering each option and f is a known function (see Batchelor and Orr, 1988, p. 322, formula (11)) and $\tilde{\mu}_t$ is the current perceived inflation rate that has to be specified.

We experimented with three choices of $\tilde{\mu}_t$: (i) recursive Hodrick–Prescott filter, (ii) past inflation, (iii) inflation extracted as fitted values from the regression of inflation on the past balance statistics from the survey. The recursive HP filter was calculated using the following quasi-real-time procedure to minimize the well-known end-of-sample problems. For each period, t , we first forecast the underlying inflation process for the next twelve quarters with an ARMA model, selected with the Akaike criterion (with the maximum number of four lags on both AR and MA terms). We then apply the filter on this artificially extended series (with the HP filter with the usual quarterly penalty

parameter $\lambda_{HP} = 1600$). Finally, we set $\tilde{\mu}_t$ equal to the value of the HP filtered inflation as of time t .

Most results reported in the paper hold for all normalizations considered. In particular, we find that the alternative rescaling procedures typically imply that λ_1 is positive and significant and the summing-up restriction $\lambda_1 + \lambda_2 = 1$ is met (using the same notation as below). One result that does not hold for normalizations (ii) and (iii) is that household expectations are insensitive to past inflation. This is not surprising because for these two normalizations past inflation rate directly enters $\tilde{\mu}_t$.

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