A Model for Forecasting
Swedish Inflation

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Summary
We apply a “P-Star” model to explain the inflation process in Sweden for the period of the early 1980s to the beginning of 2004. We provide empirical considerations on the stability of the demand for a “broadly defined” money function in Sweden, being a precondition for money having a reliable impact on (future) inflation. We find that money – when measured by P-star or, alternatively: the “real money gap” – plays an important role for explaining inflation in Sweden. Our results might thus suggest that money should play a (more) prominent role in the Riksbank’s policy making compared to the status quo.

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1. Introduction

After being forced to withdraw from the European Exchange Rate Mechanism (ERM) in November 1992, the Riksbank’s Governing Board decided in January 1993 to adopt an explicit “inflation targeting” (IT) regime (Heikensten and Vredin (2002), p. 8).\(^1\) The bank announced to keep “headline” inflation, measured as the change in the consumer price index, to be limited to 2 per cent, with a tolerance of ±1 percentage point. The level of inflation targeted by the Riksbank corresponds to that chosen by most other central banks that aim monetary policy at price stability, especially those in the European Union at that time. Also, the bank chose a headline inflation measure as its target since the public was familiar with such a measure. The main purpose of the tolerance interval was to indicate that deviations from the target level would probably occur and, at the same time, emphasise the Riksbank’s ambition to restrict such deviations.

IT has become a prominent monetary policy concept for central banks in a number of countries since the early 1990s. For instance, New Zealand adopted such an approach in 1989, Canada in 1991, the UK and Finland in 1992. Under IT, which has often been characterised as an antipode to monetary targeting (MT), the central bank bases its interest rate decision on deviations of the bank’s inflation forecast from its (explicitly) pre-announced inflation target. That is, the inflation forecast is actually the key variable for monetary policy making. This leads, of course, to the classical question about the factors actually driving (future) inflation: Is inflation a “demand pull” or “cost push” phenomenon in the Keynesian sense? Or is inflation, according to the dictum of Milton Friedman, “always and everywhere a monetary phenomenon”?

As of today, many central banks generally hold the view that money is of little use as

\(^1\) After allowing for a “transition period”, the framework became operational only from 1995 onwards. For a detailed discussion of how IT was put into practice in Sweden, see Svensson (1995).
an indicator for monetary policy aiming at stable and low inflation. This is largely due to the assumed instability of money demand relations in economies with well-developed financial markets. This general view is also expressed by the Riksbank, which states that “(...) technological developments, deregulation and internationalisation have all made it more difficult to utilise the quantity of money as an intermediate goal towards a final price stability goal.” (Mitlid and Versterlund (2001), p. 23). Against this background, the new theoretical approaches have analysed equilibrium inflation determination without any reference to either money supply or demand (see, for instance, Woodford (1997)).

For the period ranging from the second quarter of 1972 to the fourth quarter of 1995 Baumgartner et al. (2003) showed that narrow money M0 was the most powerful leading indicator for Swedish inflation and that broadly defined money M3 as well as inflation expectations had significant predictive information for inflation: “Both monetary aggregates contain information about inflation sufficiently far into the future to allow the policymakers to respond to this information in a meaningful way” (Baumgartner et al. (2003), p. 14). Interestingly enough, these findings did not attract much attention either in further research on the role of money for future inflation in Sweden or in the Riksbank’s actual monetary policy making. In particular, the body of empirical analyses on the stability of money demand – which is a prerequisite for money having a reliable influence on (future) inflation – is actually scarce to say the least.3

In what follows, we set out the major characteristics of the Riksbank’s IT concept. In

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2 It is interesting to note in the current environment of “no money” in monetary policy that Söderström (2001) argues that giving a prominent role to money can be a sensible strategy for a central bank pursuing IT under discretion as it would make discretionary policy more inertial.

3 For discussion of the difficulties of sustaining, inter alia, monetary targets as the strategy for controlling inflation, see Svensson (1994) or Obstfeld and Rogoff (1995).
this context we will also highlight conceptual and theoretical criticism which can be levied against IT as such. In a second step we will outline the “P-Star” model, which will form the starting point of explaining the Swedish inflation process. This analysis actually includes empirical considerations about the stability of the money demand function in Sweden as the latter is a precondition for money having a reliable impact on (future) inflation. In this context we also provide the model’s “out-of-sample” forecasts for inflation. The analysis concludes with a summary of the findings and an outlook on future challenges under a policy framework of IT.

2. The Riksbank’s inflation targeting concept

Bernanke et al. (1999) described IT the following way: “Inflation targeting is a framework for monetary policy characterised by the public announcement of official quantitative targets (or target ranges) for the inflation rate over one or more time horizons, and by explicit acknowledgement that low, stable inflation is monetary policy's primary long-run goal. Among other important features of inflation targeting are vigorous efforts to communicate with the public about the plans and objectives of the monetary authorities, and, in many cases, mechanisms that strengthen the central bank's accountability for attaining those objectives.”

In line with this widely accepted definition, the Riksbank has shown an increasing desire to formulate explicit and increasingly precise objectives for monetary policy, in particular numerical inflation targets (Heikensten and Vredin (2002), p.6). Moreover, steps have been taken to create an institutional setting that makes the central bank strongly committed to its objectives, in particular making the central bank politically

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4 See also Svensson (1999).
independent. Also, the central bank has developed particular decision making processes in which inflation forecasts play a very important role. Finally, the Riksbank has taken great efforts to render its policy more transparent because this makes it easier for the public to understand the central bank’s actions. An explicit objective also makes it easier for the public at large to evaluate monetary policy and hold the central bank accountable for its decisions, thereby making policy more credible from the point of view of market agents.

As far as the policy objective is concerned, the Riksbank makes, in addition to headline inflation, use of various “core inflation” measures that exclude certain price components such as, for instance, indirect taxes and subsidies as well as house mortgage interest payments. In fact, the bank has decided not to use one specific core index for all situations. The intention instead is always to communicate exactly on what forecast or other grounds the interest rate decision has been based and which deviations from the target are acceptable in any given situation because of temporary supply shocks. Figure 1 summarizes the Riksbank’s inflation measures. Figure 2 shows the latest developments of Swedish inflation measures. As can be seen, inflation has declined in recent years, developing more or less closely aligned to the bank’s stability promise.

In general there is agreement that a pre-emptive, forward-looking monetary policy such as IT shall take action if and when there is a divergence between expected, or projected ($\pi_t^e$) and envisaged ($\hat{\pi}_t$) inflation. The policy recommendation under an IT framework could be described as follows:

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5 The new Constitution and amended Riksbank Act came into effect in 1999 granting political independence to the central bank. The Riksbank Act now states that the “objective of the Riksbank’s operations shall be to maintain price stability.” See Heikensten and Vredin (2002, p. 9).
The central bank would have to increase (decrease) the interest rate, \( i \), that is \( \Delta i > 0 \) \((\Delta i < 0)\), if expected future inflation exceeds target inflation; \( \lambda^* > 0 \) shows the intensity with which rates are changed in response to the expected deviation from target inflation.

From the point of view of monetary policy, it seems advisable to analyse risks to price stability by taking into account both monetary and non-monetary variables. Such an analysis would combine Milton Friedman’s dictum that “inflation is always and everywhere a monetary phenomenon” with the fact that consumer prices are also temporarily influenced by “cost push” and “demand pull” variables (for instance, variations in the output gap, the oil price, the wage level and the exchange rate).

In this context it should be noted that conceptually, IT and MT are much more closely aligned than most discussions would suggest.\(^6\) Both concepts aim to keep (future) inflation in check; both favour a pre-emptive stance for monetary policy; and both favour policy-making on the basis of inflation forecasts. MT proponents would argue for using money supply as the central inflation indicator, whereas those in favour of IT recommend a central bank’s “self-made” inflation forecast as the main guideline for policy-making. That is, MT and IT would be identical if money supply were used as the inflation forecast variable. The only difference remaining in such a case would be that MT has an explicitly announced money growth target and an implicit inflation goal, whereas IT has an explicit inflation target and an implicit money growth goal. In view of the above, it is fair to say that IT could be characterised as an “umbrella strategy” under which money supply and other variables can be analysed in order to identify risks for future price stability.

\(^6\) For an insightful comparison between MT and IT, see, for instance, Baltensperger (2000).
Under IT, a central bank’s inflation forecast plays the key role in policy making. However, it is not quite clear how such forecasts are actually calculated. Of course, in its quarterly Inflation Reports the Riksbank provides – along with its inflation forecasts one to two years ahead, and nowadays there is also a brief outlook three years ahead – some input variables for its forecasts. At the same time, however, it is not at all clear which role each variable plays, e.g. which weight is assigned to it in the Riksbank’s forecasting exercise. In fact, its inflation forecasts – as is the case with other central bank pursuing IT – appear to emerge from a kind of “black box”, which would imply a great deal of discretion on the part of the forecasters. This conjecture is actually confirmed by the bank: “(…) the Riksbank’s forecasts (like those of other central banks) are largely determined by judgement.” (Jansson and Vredin (2001), p. 206).

A wide scope of discretion in calculating the Riksbank’s IT inflation forecast might not only raise questions about the actual “transparency” of its monetary policy: if the bank derives the essential policy making variable on an “ad hoc” basis and under a rather opaque procedure, it could well be that in a “period of stress” (such as, for instance, big financial market and/or cost push crises) the public looses confidence in the bank’s assessment. (Interestingly enough, even the most vigorous advocates of central bank policy transparency seem to be satisfied with the actual status quo.) It might also raise questions about the actual empirically reliable relation between (future) inflation and its “driving forces”. That is, we try to shed some more light on the actual role money plays in the inflation process in Sweden. To this end, we will make use of the well-known “P-

7 “The discussion of Swedish inflation is more or less based on an expectations-augmented Phillips curve framework. Thus, supply and demand conditions in the Swedish economy are discussed along with various measures of the “output gap”, and the picture of inflation is elaborated with inflation expectations and possible supply shocks.” (Heikensten and Vredin (2002), p. 16). It is interesting to note in this context that Baumgartner et al. (2003, p. 5) note: "(…) the output gap (…) has some predictive information on inflation, but the predictive information of the output gap is confined to a shorter horizon than the monetary aggregates (…)."
star” model.

3. The “P-Star” model

To set the ball rolling, Figure 4 shows the annual rise in consumer price inflation (left hand scale) and the income velocity of the stock of M3 (right hand scale) in Sweden for the period Q1 85 to Q1 04. As can be seen, there is a certain co-movement between the swings of inflation and the behavior of the velocity of money: rising (falling) inflation seem to be associated with upwards (downward) swings in the velocity of money. That is, “swings” in the velocity of money or, to put it differently, swings in the demand for money, seem to be accompanied by movements in inflation. This finding, however, is by no means “spurious” when taken into account the well-known “transaction equation” and – on the basis of this – the so-called “P-star” model. We will now look into this in more detail.

- Insert Figure 4 here -

For setting up an inflation forecasting model for Sweden, we make use of the well-known “transaction equation” which can be written as follows:

\[ M \cdot V = Y \cdot P \, , \]  

(1)

where \( M \) is the stock of money, \( V \) the velocity of money, \( Y \) real output and \( P \) price level. Equation (1) simply says that the stock of money, multiplied by the number of times a money unit is used for financing purposes, equals the real output valued with its price level. Taking logarithms, equation (1) can easily be written as:

\[ \ln M + \ln V = \ln Y + \ln P \, . \]  

(2)

Now let us turn to the “P-star” model (Hallman, Porter and Small (1991)). To start with,

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8 The velocity of money is the frequency with which a money unit is used to finance nominal output in the period under review. Economically speaking, the velocity of money is the reciprocal of the demand for money. It is simply calculated by dividing nominal GDP by the stock of money.
the actual price level is simply:

\[ p = m + v - y . \]  

(3)

The long-term price level can be formalised as:

\[ p^* = m + v^* - y^* , \]  

(4)

where asterisks represent the long-run or equilibrium values. The difference between equations (4) and (3) is the so-called price gap:

\[ p^* - p = (v^* - v) + (y - y^*). \]  

(5)

The price gap \( p^* - p \) consists of (i) the liquidity gap \( v^* - v \) and (ii) the output gap \((y^* - y)\). If, for instance, actual output exceeds potential \((y^* < y)\) and actual velocity equals the long-term equilibrium \((v = v^*)\), the actual price level can be expected to rise in the future. The price gap can also be written as:

\[ \left( p^*_t - p_t \right) = \left( m_t + v^*_t - p_t \right) - y^*_t . \]  

(6)

According to equation (5), the price gap is a function of the “output gap”, that is the difference between actual and potential GDP, and the “liquidity gap”, defined as the difference between the equilibrium velocity of money and its actual value. It is important to note that an increase in real GDP \((y)\) will not cause a change of the price gap, because \(v\) will decrease as \(y\) increases. Equation (6) shows that the price gap is independent from the output gap: it is simply the difference between real money (adjusted by the trend velocity) and real potential output. Figure 5 shows the output gap and the price gap for Sweden.

- Figure 5 about here -

Lately, there is an alternative version of the price gap, namely the “real money gap” (Gerlach and Svensson (2001)). It is defined as actual money supply less actual price
level:

\[ m_{\text{real}} = m - p. \]  

(7)

The equilibrium real money holding is:

\[ m'_{\text{real}} = m - p^*. \]  

(8)

The difference between equation (7) und (8) is the real money gap, which represents nothing other than the price gap with a negative sign:

\[ m_{\text{real}} - m'_{\text{real}} = (m - p) - (m - p^*) = -p + p^* = -(p - p^*). \]  

(9)

The real money gap is thus very closely affiliated with the so-called P-star model. It is crucial to identify a stable velocity of money or, equivalently, a stable demand function for money. Only when there is a stable demand for money, monetary signals might contain information for future inflation. This issue will be dealt with what follows next.

4. Identifying a stable demand function for money

Theory suggests that holdings of real balances depend, in the simplest case, on output and interest rates, that is opportunity costs of money holdings. An increase (decline) in output should be accompanied with rising (falling) demand for transaction balances. As a result, the income elasticity of money should be positive. The higher (lower) the opportunity costs of real money holdings are, the lower (higher) should be the demand for real balances. Also, a rise (decline) in the own rate on deposits included in the monetary aggregate can be expected to increase (reduce) the demand for real money. A parsimoniously specified type money demand can thus be written as follows:

\[ \text{See also Appendix A. Against the background of these findings, it is easily shown that a simple comparison between actual money growth and an annual growth rate (such as, for instance, represented by the ECB’s reference value concept) might lead to misleading policy signals as monetary expansions, which occurred in the past and will have a bearing on future prices, are systematically neglected.} \]
\[
\begin{align*}
m_t - p_t &= \beta_0 + \beta_1 y_t + \beta_2 \Delta_{\text{short}} + \beta_3 \Delta_{\text{long}} + \epsilon_t \\
\end{align*}
\]  

(10)

where \( m \) is the logarithm of money, \( p \) is the price level, \( y \) represents real GDP and \( i \) are interest rates, and \( \epsilon \) is the error term. The \( \beta \)s are the parameter to be estimated.

To analyze the stability of money demand functions, the cointegration analysis suggested by Johansen (1988, 1991) and Johansen and Juselius (1992) has become the standard econometric approach (see Appendices C to E for further details). Without going into detail at this juncture, the rationale of the procedure is to identify linear relationships between non-stationary variables, in our case real money, output, interest rates and inflation. Moreover, our multivariate approach can be subject to various restrictions. For instance, we assumed that there are linear combinations between (i) real M3 holdings, real GDP and interest rates and (ii) short-term interest rates, long-term rates and inflation. The first assumption actually implies the existence of a demand for money function, the second assumes that the “term structure of interest rates” and the “Fisher parity” jointly hold. Our approach therefore amounts to a structural vector error correction model (S-VECM), a special case of the structural vector auto regression model (S-VAR). The cointegration analysis for the period Q1 84 to Q1 04 reveals the following results:

\[
\begin{bmatrix}
\Delta y_t \\
\Delta i_{\text{long}} \\
\Delta i_{\text{short}} \\
\Delta \text{cpi}_t
\end{bmatrix} = 
\begin{bmatrix}
1 & 0 & -1.35 & 0.25 & 0 & 0 \\
0 & 0 & 0 & 1 & -10.7 & 1 \\
0 & 0 & 0 & 1 & -0.11 & 0 \\
\end{bmatrix}
\begin{bmatrix}
m_t - p_t \\
y_t \\
i_{\text{long}} \\
i_{\text{short}} \\
\Delta \text{cpi}_t
\end{bmatrix}
\]  

(11)

The first vector represents the demand for money function. The income elasticity of money is 1.31, which is economically plausible and broadly in line with other studies.\(^\text{10}\)

\(^{10}\)The finding that the income elasticity exceeds 1 can be explained by the „omitted wealth effect”, which leads to a decline of the velocity of money over time.
The demand for balances rises (declines) when the opportunity costs declines (rises). Also, the demand for real money increases (declines) if the own yield on money deposit increases (declines). The second vector actually implies that the term structure of interest rates and the Fisher parity hold (that is that they are jointly stationary). We also tested the system using a measure of core inflation, $\Delta \text{cpicore}_t$. (For the co-movement of headline and core inflation see Figure 6.) The analysis yields the following results, which are closely related to the system using headline inflation:

$$
\hat{\beta}'X_i = \begin{bmatrix}
1 & -1.43 \\
0 & 0 \\
0 & -15.3 \\
0 & 1 -0.19
\end{bmatrix}
\begin{bmatrix}
m_t - p_t \\
y_t \\
ig_t \\
\Delta \text{cpicore}_t
\end{bmatrix}
$$

(12)

We take these findings as sufficient to consider the demand for M3 function as reasonably stable. That is, we will set up an inflation forecasting model using the price gap as an inflation determining factor.

5. Inflation estimating model and out-of-sample forecasts

In view of the findings above, our inflation forecasting model describes the dynamics of inflation as follows:

$$
\pi_{t+1} = \beta_0 + \beta_1 (p_{t*} - p_t) + \sum_{i=1}^{n} \beta_i \pi_{t-i} + N_t + \varepsilon_t
$$

(13)

where $\pi_{t+1}$ is future inflation. $p_{t*} - p_t$ is the price gap (see equation (6)). If the actual price level is lower (higher) than the equilibrium level, future inflation will accelerate (slow down) to close the “gap”. As a result, one would expect the parameter $\beta_i$ to be positive. Given the “stickiness” of inflation, we also take into account past inflation as
shown by \( \sum_{i=1}^{n} \beta_i \pi_{t,i} \). \( N_{t} \) represents a vector of non-monetary “cost push” variables (oil, wages, exchange rates, unemployment etc.). Finally, the term \( \epsilon_t \) is the (i.i.d.) error term.

We regressed inflation – expressed as the first difference of the log consumer price index – on own lagged values (for doing justice to the “stickiness” of the inflation process), and (first changes in the logs of) various variables, namely the price gap (\( pg \)), output gap (\( og \)), oil prices (\( oil \)), the USDSEK exchange rate (\( exsek \)), unemployment rate (\( u \)) and nominal wages (\( w \)). Figure 7 shows the regression results and a selection of diagnosis tests for the period Q3 84 to Q1 04. The equation explains 79% of the variance in the change of inflation and appears to be stable according to standard tests. This finding is also confirmed when various sub-periods are tested. For instance, Figure 8 shows an out-of-sample calculation where the parameters were established for the (sub-)period Q3 84 to Q4 95.

It is interesting to note that the price gap enters the forecasting model with a time lag of five quarters whereas the output proved to be insignificant according to standard tests. That said, the past (current) values of the changes in the price gap exert a rather strong impact on current (future) inflation. Given the latest benign development of the price gap, there is only very moderate upward pressure on the consumer price level around the beginning of 2004. Figure 9 shows impulse-response functions for real money holdings and real output due to a change in the interest rate and the response of the long-term interest rate and inflation following a shock to the real money supply. The variables under review reveal economically plausible reactions.
It is possible for an equation that has a very good statistical fit to have a very poor simulation or forecast fit. For this reason, we finally make use of both static and dynamic out-of-sample forecasts. Static forecasting performs a series of one-step ahead forecasts of the dependent inflation variable, while dynamic forecasting relies on a multi-step forecast of inflation. Both methods yield identical results for the first forecast period. However, the two methods will differ for subsequent periods since there are lagged dependent variables included in our empirical model. Let us start now with the static out-of-sample forecast which starts in the first quarter of 1996.

Static forecasts

Figure 10 plots the inflation forecasts with plus and minus two standard error bands which provide a 95% forecast interval and report a table of statistical results evaluating our static forecast. The first two forecast error statistics depend on the scale of the dependent variable and hence should be used to compare forecasts for the inflation rate across different models. The popular Root Mean Squared Error (RMSE) turns out to be small and implicitly weights large forecast errors more heavily than small ones and is appropriate in situations in which the cost of an error increases as the square of that error. In addition, the mean absolute deviation is rather small. This appears to be more important in our IT context since the cost of inflation forecast errors is proportional to the absolute size of the forecast error and not so much to the percentage error as implicitly assumed by the Mean Absolute Percentage Error (MAPE) (see, for instance, Kennedy (2003), pp. 361).

However, we emphasize the remaining two statistics which are scale invariant. The Theil inequality coefficient always lies between zero and one, where zero indicates a perfect fit. Our forecast appears to be rather good since Theil falls below 50% and the
bias and variance proportions are so small that most of the bias is concentrated on the covariance proportions. Hence, the mean of the forecast is not far away from the mean of the actual series. The same is valid for the variation of our forecast (see, e.g., Pindyck and Rubinfeld (1991), chapter 12). Also the MAPE turns out to be small. However, note that a well-known problem connected with this indicator is that it promotes a kind of underestimating due to the fact that the actual and not the forecasted value is the base for calculating the percentage error. Figure 11 shows the results.

- Insert Figure 11 here -

**Dynamic Forecast**

The first forecasted quarter of our dynamic forecast is Q1 96. Figure 12 plots the inflation forecasts with plus and minus two standard error bands which provide a 95% forecast interval and report a table of statistical results evaluating our dynamic forecast.

- Insert Figure 12 here -

Again, like in the static forecast case, the forecast evaluation leads to a rather positive assessment of our empirical model. The RMSE and the MAPE turn out to be small again. The Theil inequality coefficient is nearly unchanged in spite of our change from a static to a dynamic forecast perspective. However, the bias and variance proportions are slightly higher than before. Hence, the mean and the variance of the forecast are more far away from the mean and the variation of the actual series, a finding typical for and inherent in dynamic forecasts. Finally, also the MAPE turns out to be small, and thus acceptable. Figure 13 shows the actual and dynamically forecasted inflation.

- Insert Figure 13 here -
6. Summary and outlook

Applying the P-star model to Swedish data from the early 1980 to 2004 suggests that broadly defined money, measured by the concept of P-star or real money gap, plays an important role in explaining Swedish inflation. This finding rests on the observation that the demand for money function, e.g. the income velocity of money, appears to be stable in the long-run. Our findings are supported by statistically satisfactory out-of-sample inflation forecasts based on the P-star approach. Our findings on the role of money for future inflation may come as a surprise as the Riksbank has not assigned – at least explicitly – any important role to the stock of money in its monetary policy making. Our findings may thus provide a basis from which further research on whether money should in fact be assigned an intermediate function in the Riksbank’s policy making could be done.

Moreover, our findings may also be taken as a starting point to discuss some “weak spots” in the Riksbank’s IT approach. Perhaps most importantly in this context, it should be mentioned that – despite the successful policy results in recent years – there is little evidence that could inspire confidence in the Riksbank’s inflation forecasts playing an important role of determining future inflation. The well-known phenomenon of “surprise inflation” – that is actual inflation in excess of market agents’ originally expected inflation – attests to this. The role of excess money as measured by the real money gap, in turn, provides a theoretically and empirically sound explanation of the policy transmission mechanism linking monetary policy action with the ultimate policy target, that is consumer price inflation.

To the outside world, inflation forecasts are rather opaque: it is not quite clear which variables are included in the projection model; nor is it known how much weight is assigned to each of the variables. So the public’s confidence in the accuracy of the
inflation projections – and the appropriateness of its policy recommendations – can be assumed to hinge de facto on the bank’s credibility, that is the bank’s perceived willingness and ability to deliver on its price stability promise. It therefore seems questionable whether inflation forecasts themselves further monetary policy transparency and build up central bank credibility. It seems to work more the other way round: Inflation projections (or forecasts) are only reliable if central bank credibility is already in place.

Finally, IT has gained in prominence in a period in which world wide inflation was set to decline – for reasons that might have been well beyond the factors considered important in the traditional IT approach. For instance, the political, financial and operational independency granted to numerous central banks – including the Riksbank – should have played a role in making monetary policy credible from the point of view and bringing inflation down to the desired levels. In the Swedish case, though, our findings show in particular that the decline in the real money gap seems to be at the heart of the disinflation process, i.e. in bringing inflation to the envisaged level. This result would indeed suggest that the Riksbank might have to assign a (more) prominent role to money in setting its policy compared to the status quo.
References


Appendix

A The price gap

To derive P-star, a simple demand for money function may be used:

\[ m = p + \beta_1 y - \beta_2 i + \varepsilon , \]  

(A1)

where \( m \) is the (log) of the stock of nominal money, \( p \) the price level and \( i \) the interest rate, e.g. the opportunity costs of holding money. The parameter \( \beta_1 \) represents the income elasticity of money, \( \beta_2 \) is the (semi-)elasticity of money demand. The term \( \varepsilon \) represents the error term. In the long run, or equilibrium, nominal money holdings can be described as follows:

\[ m^* = p + \beta_1 y^* - \beta_2 i^* , \]  

(A2)

whereas the asterisks stand for long-term, that is equilibrium, values. The difference between the current money stock and the equilibrium money stock is the “money gap”:

\[ m - m^* = \beta_1 (y - y^*) - \beta_2 (i - i^*) + \varepsilon . \]  

(A3)

Instead of measuring these disequilibria in units of the (logarithmic) money stock, they can also be expressed in an equivalent manner in units of the (logarithmic) price level. To do this, the equilibrium price level (P-star) is defined as the price level that would emerge given the current holdings of money if both the goods market and the money market were in equilibrium:

\[ p^* = m - \beta_1 y^* + \beta_2 i^* . \]  

(A4)

The equilibrium price level is thus an indicator of the level of goods prices that would emerge over the longer term given the existing money stock if the disequilibria \((y - y^*, i - i^*, \varepsilon)\) had disappeared. As may easily be seen, the price gap and the money gap are identical:
\[ p^* - p = \beta_1(y - y^*) - \beta_2(i - i^*) + \varepsilon, \]  
(A5)

so that we can write:

\[ p^* - p = (m^* - m) + \varepsilon \]  
(A6)

B Data

Nominal and real GDP, consumer prices, interest rates, the oil price and exchange rates were taken from Thomson Financials. The stock of M3 was taken from the Riksbank and was seasonally adjusted (Census X12). Core consumer prices were taken from the OECD. Except for interest rates, all variables are seasonally adjusted and were used in the form of natural logarithms.

C Overview of the Johansen cointegration approach

The finding that many macro time series may contain a unit root has spurred the development of the theory of non-stationary time series analysis. Engle and Granger (1987) pointed out that a linear combination of two or more non-stationary series may be stationary. If such a stationary linear combination exists, the non-stationary time series are said to be cointegrated. The stationary linear combination is called the cointegrating equation and may be interpreted as a long-run equilibrium relationship among the variables.

The purpose of the cointegration test is to determine whether a group of non-stationary series are cointegrated or not. As explained below, the presence of a cointegrating relation forms the basis of the vector error correction (VEC) specification. To outline the Johansen (1991, 1995) cointegration technique, consider a vector autoregressive (VAR) model of the order \( p \):

\[ y_t = A_1 y_{t-1} + \ldots + A_p y_{t-p} + Bx_t + \varepsilon_t \]  
(A1)
where $y_t$ is a $p$-vector of non-stationary $I(1)$ variables, $x_t$ is a $d$-vector of deterministic variables, and $\varepsilon_t$ is a vector of innovations. We can rewrite this VAR as:

$$\Delta y_t = \Pi y_{t-1} + \sum_{i=1}^{p-1} \Gamma_i \Delta y_{t-i} + Bx_t + \varepsilon_t$$  \hspace{1cm} (A2)

where

$$\Pi = \sum_{i=1}^{p} A_i - I \text{ and } \Gamma_i = - \sum_{j=i+1}^{p} A_j .$$  \hspace{1cm} (A3)

Granger’s representation theorem asserts that if the coefficient matrix $\Pi$ has reduced rank $r < k$, then there exist $k \times r$ matrices $\alpha$ and $\beta$ each with rank $r$ such that $\Pi = \alpha \beta^\prime$ and $\beta^\prime y_t$ is $I(0)$. $r$ is the number of cointegrating relations (the rank) and each column of $\beta$ is the cointegrating vector. The elements of $\alpha$ are known as the adjustment parameters in the VEC model. Johansen's method is to estimate the $\Pi$ from an unrestricted VAR and to test whether we can reject the restrictions implied by the reduced rank of $\Pi$.

To determine the number of cointegrating relations conditional on the assumptions made about the trend, one can proceed sequentially from $r = 0$ to $r = k - 1$ until one fails to reject.

The trace statistic reported in the first block tests the null hypothesis of $r$ cointegrating relations against the alternative of $k$ cointegrating relations, where $k$ is the number of endogenous variables, for $r = 0, 1, \ldots, k - 1$. The alternative of $k$ cointegrating relations corresponds to the case where none of the series has a unit root and a stationary VAR may be specified in terms of the levels of all of the series. The trace statistic for the null hypothesis of $r$ cointegrating relations is computed as:

$$LR_u (r|k) = -T \sum_{i=r+1}^{k} \log(1-\lambda_i)$$  \hspace{1cm} (A4)
where $\lambda_i$ is the $i$-th largest eigenvalue of the $\Pi$ matrix in (A3) which is reported in the second column of the output table. The maximum eigenvalue statistic shows the results of testing the null hypothesis of $r$ cointegrating relations against the alternative of $r + 1$ cointegrating relations. This test statistic is computed as:

$$LR_{\max}(r|r + 1) = -T \log(1 + \lambda_{r+1})$$

$$(A5)$$

$$= LR_{\max}(r|k) - LR_{\max}(r|r + 1)$$

for $r = 0, 1, \ldots, k - 1$. 

### D Cointegration Analysis, Headline Inflation

Table A2: Cointegration rank tests

#### Unrestricted Cointegration Rank Test

<table>
<thead>
<tr>
<th>Hypothesized</th>
<th>Trace Statistic</th>
<th>5 Percent</th>
<th>1 Percent</th>
</tr>
</thead>
<tbody>
<tr>
<td>No. of CE(s)</td>
<td>Eigenvalue</td>
<td>Critical Value</td>
<td>Critical Value</td>
</tr>
<tr>
<td>None **</td>
<td>0.593715</td>
<td>140.7056</td>
<td>68.52</td>
</tr>
<tr>
<td>At most 1 **</td>
<td>0.394824</td>
<td>67.74881</td>
<td>47.21</td>
</tr>
<tr>
<td>At most 2</td>
<td>0.185623</td>
<td>27.06776</td>
<td>29.68</td>
</tr>
<tr>
<td>At most 3</td>
<td>0.119402</td>
<td>10.43590</td>
<td>15.41</td>
</tr>
<tr>
<td>At most 4</td>
<td>0.001682</td>
<td>0.136394</td>
<td>3.76</td>
</tr>
</tbody>
</table>

*(**) denotes rejection of the hypothesis at the 5%(1%) level

Trace test indicates 2 cointegrating equation(s) at both 5% and 1% levels

#### Hypothesized Max-Eigen Rank Test

<table>
<thead>
<tr>
<th>Hypothesized</th>
<th>Max-Eigen Statistic</th>
<th>5 Percent</th>
<th>1 Percent</th>
</tr>
</thead>
<tbody>
<tr>
<td>No. of CE(s)</td>
<td>Eigenvalue</td>
<td>Critical Value</td>
<td>Critical Value</td>
</tr>
<tr>
<td>None **</td>
<td>0.593715</td>
<td>72.95678</td>
<td>33.46</td>
</tr>
<tr>
<td>At most 1 **</td>
<td>0.394824</td>
<td>40.68105</td>
<td>27.07</td>
</tr>
<tr>
<td>At most 2</td>
<td>0.185623</td>
<td>16.63186</td>
<td>20.97</td>
</tr>
<tr>
<td>At most 3</td>
<td>0.119402</td>
<td>10.29951</td>
<td>14.07</td>
</tr>
<tr>
<td>At most 4</td>
<td>0.001682</td>
<td>0.136394</td>
<td>3.76</td>
</tr>
</tbody>
</table>

*(***) denotes rejection of the hypothesis at the 5%(1%) level

Max-eigenvalue test indicates 2 cointegrating equation(s) at both 5% and 1% levels
Table A3: Tests of hypotheses about the conintegration space

Cointegration Restrictions:

\[ B(1,1) = 1, B(1,3) = 0, B(1,5) = 0 \]
\[ B(2,1) = 0, B(2,2) = 0, B(2,4) = 1 \]

Maximum iterations (500) reached.
Restrictions identify all cointegrating vectors

LR test for binding restrictions (rank = 2):

| Chi-square(2) | 4.781542 |
| Probability   | 0.091559 |

<table>
<thead>
<tr>
<th>Cointegrating Eq:</th>
<th>CointEq1</th>
<th>CointEq2</th>
</tr>
</thead>
<tbody>
<tr>
<td>LNM3(-1)-LNCPI(-1)</td>
<td>1.000000</td>
<td>0.000000</td>
</tr>
<tr>
<td>LNGDPR(-1)</td>
<td>-1.355166</td>
<td>0.000000</td>
</tr>
<tr>
<td></td>
<td>(0.19241)</td>
<td>[-7.04301]</td>
</tr>
<tr>
<td>SW10Y(-1)</td>
<td>0.000000</td>
<td>-10.78579</td>
</tr>
<tr>
<td></td>
<td>(1.16628)</td>
<td>[-9.24806]</td>
</tr>
<tr>
<td>SW3M(-1)</td>
<td>0.246110</td>
<td>1.000000</td>
</tr>
<tr>
<td></td>
<td>(0.03906)</td>
<td>[ 6.30003]</td>
</tr>
<tr>
<td>D(LNCP1(-1))</td>
<td>0.000000</td>
<td>-106.0225</td>
</tr>
<tr>
<td></td>
<td>(80.3483)</td>
<td>[-1.31954]</td>
</tr>
<tr>
<td>C</td>
<td>16.57327</td>
<td>1.098341</td>
</tr>
</tbody>
</table>
### Cointegration Analysis, Core Inflation

Table A4: Cointegration rank tests

**Unrestricted Cointegration Rank Test**

<table>
<thead>
<tr>
<th>Hypothesized</th>
<th>Trace Eigenvalue</th>
<th>5 Percent</th>
<th>1 Percent</th>
</tr>
</thead>
<tbody>
<tr>
<td>None **</td>
<td>0.576955</td>
<td>139.4359</td>
<td>68.52</td>
</tr>
<tr>
<td>At most 1 **</td>
<td>0.410480</td>
<td>69.75338</td>
<td>47.21</td>
</tr>
<tr>
<td>At most 2</td>
<td>0.178335</td>
<td>26.94924</td>
<td>29.68</td>
</tr>
<tr>
<td>At most 3</td>
<td>0.127339</td>
<td>11.03905</td>
<td>15.41</td>
</tr>
<tr>
<td>At most 4</td>
<td>7.61E-05</td>
<td>0.006161</td>
<td>3.76</td>
</tr>
</tbody>
</table>

*(**) denotes rejection of the hypothesis at the 5%(1%) level
Trace test indicates 2 cointegrating equation(s) at both 5% and 1% levels

<table>
<thead>
<tr>
<th>Hypothesized</th>
<th>Max-Eigen Eigenvalue</th>
<th>5 Percent</th>
<th>1 Percent</th>
</tr>
</thead>
<tbody>
<tr>
<td>None **</td>
<td>0.576955</td>
<td>69.68248</td>
<td>33.46</td>
</tr>
<tr>
<td>At most 1 **</td>
<td>0.410480</td>
<td>42.80413</td>
<td>27.07</td>
</tr>
<tr>
<td>At most 2</td>
<td>0.178335</td>
<td>15.91020</td>
<td>20.97</td>
</tr>
<tr>
<td>At most 3</td>
<td>0.127339</td>
<td>11.03289</td>
<td>14.07</td>
</tr>
<tr>
<td>At most 4</td>
<td>7.61E-05</td>
<td>0.006161</td>
<td>3.76</td>
</tr>
</tbody>
</table>

*(**) denotes rejection of the hypothesis at the 5%(1%) level
Max-eigenvalue test indicates 2 cointegrating equation(s) at both 5% and 1% levels
Table A5: Tests of hypotheses about the cointegration space

Cointegration Restrictions:

\[ B(1,1)=1, B(1,3)=0, B(1,5)=0 \]
\[ B(2,1)=0, B(2,2)=0, B(2,4)=1 \]

Maximum iterations (500) reached.

Restrictions identify all cointegrating vectors

LR test for binding restrictions (rank = 2):

| Chi-square(2) | 3.435574 |
| Probability   | 0.179463 |

<table>
<thead>
<tr>
<th>Cointegrating Eq:</th>
<th>CointEq1</th>
<th>CointEq2</th>
</tr>
</thead>
<tbody>
<tr>
<td>LNM3(-1)-LNCPIC(-1)</td>
<td>1.000000</td>
<td>0.000000</td>
</tr>
<tr>
<td>LNGDPR(-1)</td>
<td>-1.431439</td>
<td>0.000000</td>
</tr>
<tr>
<td></td>
<td>(0.19681)</td>
<td>[-7.27321]</td>
</tr>
<tr>
<td>SW10Y(-1)</td>
<td>0.000000</td>
<td>-15.3001</td>
</tr>
<tr>
<td></td>
<td>(1.69988)</td>
<td>[-9.00062]</td>
</tr>
<tr>
<td>SW3M(-1)</td>
<td>0.258439</td>
<td>1.000000</td>
</tr>
<tr>
<td></td>
<td>(0.03795)</td>
<td>[6.81002]</td>
</tr>
<tr>
<td>D(LNCPIC(-1))</td>
<td>0.000000</td>
<td>-198.5671</td>
</tr>
<tr>
<td></td>
<td>(128.957)</td>
<td>[-1.53979]</td>
</tr>
<tr>
<td>C</td>
<td>16.61850</td>
<td>2.006364</td>
</tr>
</tbody>
</table>
Figure 1: Swedish inflation measures

**Consumer price index (CPI):** The change in the CPI measures how the prices of goods and services for private consumption develop over time.

**Underlying inflation (UND1X):** CPI inflation excluding household mortgage interest expenditure and the direct effects of changes in taxes and subsidies.

**Domestic underlying inflation (UNDINHX):** UND1X excluding mainly imported goods and services.

**Imported underlying inflation (UNDIMPX):** UND1X excluding mainly domestically produced goods and services.

**Harmonised index of consumer prices (HICP):** An EU-harmonised index that was developed to measure inflation in the EU and that enables comparisons between EU countries. A significant part of the changes in housing costs that are included in the CPI are not included in the HICP. However, the HICP includes certain components that are currently excluded from the CPI, such as charges for childcare and care of the elderly.

Figure 2: Swedish Inflation Measures, % y/y

Source: Thomson Financials, Bloomberg; own calculations.
Figure 3. – Underlying inflation (UND1X) and the Riksbank’s inflation forecasts

![Graph showing Underlying inflation (UND1X) and the Riksbank’s inflation forecasts](image)


Figure 4: Swedish Inflation and the Income Velocity of M3

![Graph showing Swedish Inflation and the Income Velocity of M3](image)

Source: Riksbank, Thomson Financials; own calculations. The income velocity of M3 is calculated by subtracting the log of M3 from the log of nominal GDP.
Figure 5: Swedish M3 Price Gap and Output Gap, % y/y

Source: Riksbank, Thomson Financials, Bloomberg; own calculations. Four quarter averages.

Figure 6: Swedish headline and core inflation, % Y/Y

Source: Riksbank, Thomson Financials; own calculations. Fourth differences of log values. Core inflation = CPI excluding energy and food.
Figure 7: Estimating Changes in Inflation, Q3 84 to Q1 04

\[
\Delta_{t,cpi} = -0.001 - 0.246 \Delta_{t-2,cpi} + 0.291 \Delta_{t-3,cpi} + 0.266 \Delta_{t-4,cpi} + 0.279 \Delta_{t-5,pgm3} \\
- 0.010 \Delta_{t,oil} + 0.217 \Delta_{t,exsek} - 0.007 \Delta_{t,DUM91} + 0.027 DUM901 - 0.029 DUM92 \\
+ 0.025 DUM931 - 0.018 DUM981
\]

\(R^2 = .79\) \ S.E. of regression = 0.0044. – LM test (2) = .34 (.71), LM test (4) = .29 (.88), LM test (8) = 1.61 (.14). – ARCH test (2) =.33 (.71), ARCH test (4) = .56 (.69), ARCH test (8) = .66 (.73). – WHITE (no cross terms) = .49 (.95) WHITE = .67 (.89).

Data source: Source: Riksbank, Thomson Financials, Bloomberg; own calculations. Legend: ln = natural logarithm, cpi = log of consumer price index, pgm3 = price gap on the basis of the stock of money M3, oil = oil price in US$, exsek = Swedish Krona US-dollar exchange rate, DUM901, DUM921, DUM 931 and DUM981 represent dummy variables which takes on the value of 1 on Q1 90, Q1 92, Q1 93 and Q1 98, respectively, and zero otherwise. All variables enter the equation in logarithms. \(\Delta_1\) represents the first difference of the log of the value. The numbers in brackets show the t-values.

Figure 8: Swedish inflation, actual and estimate (out-of-sample), % Y/Y

Source: Riksbank, Thomson Financials, Bloomberg; own calculations. Actual inflation is based on the inflation equation shown in Figure 7.
Figure 9: Impulse-response functions for the Swedish economy

Response to Cholesky One S.D. Innovations

- Response of LNM3-LNCPI to SW3M
- Response of LNGDPR to SW3M
- Response of SW10Y to LNM3-LNCPI
- Response of D(LNCPI) to LNM3-LNCPI

Source: Riksbank, Thomson Financials; own calculations.

Figure 10: Static Inflation forecasts and forecast evaluation

Forecast: DLNCPIF
Actual: D(LNCPI)
Forecast sample: 1996:1 2004:1
Included observations: 33

Root Mean Squared Error 0.004041
Mean Absolute Error 0.002911
Mean Abs. Percent Error 1.396882
Theil Inequality Coefficient 0.443157
Bias Proportion 0.012753
Variance Proportion 0.048242
Covariance Proportion 0.939005
Figure 11: Inflation – Actual and forecasted values

![Inflation Graph](image1)

Figure 12: Dynamic inflation forecasts and forecast evaluation

![Dynamic Inflation Graph](image2)

Forecast: DLNCPIF
Actual: D(LNCP)I
Forecast sample: 1996:1 2004:1
Included observations: 33

- Root Mean Squared Error: 0.004451
- Mean Absolute Error: 0.003273
- Mean Abs. Percent Error: 6.079383
- Theil Inequality Coefficient: 0.449356
- Bias Proportion: 0.156703
- Variance Proportion: 0.075484
- Covariance Proportion: 0.767813
Figure 13: Inflation – Actual and forecasted values