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Estimates of the US Phillips Curve
with the General to Specific Method

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Abstract

This paper distinguishes between the long run and short run Phillips curve (PC) and uses the micro theory based specification, with forward looking expectations, for the long run PC. The long run and the implied short run dynamic equations are estimated in one step with the general to specific method (GETS). Our approach has two distinct advantages. Firstly, classical estimation methods can be used, irrespective of the stationarity properties of the variables. Secondly, instead of arbitrarily adding the lagged inflation rate to the theory based long run PC to capture persistence in inflation, our approach shows that persistence effects can also be captured through the dynamic adjustment equations. This has an added advantage because it offers a more flexible lag structure to estimate dynamic adjustments compared to the partial adjustment process in the hybrid NKPC.

Keywords: US New Keynesian Phillips Curve, Forward looking expectations, Alternative measures of the Driving Forces, GETS.

JEL: C2, C12, E3.

1. Introduction

Recent empirical studies of the new Keynesian Phillips curve (NKPC) have discussed several issues concerning its specification and estimation. Two important controversial issues are the relative importance of the backward and forward looking expectations and whether the output gap or the share of wages is a satisfactory proxy for real marginal costs. Gali and Gertler (1999) have augmented the micro-theory based new Keynesian Phillips curve (NKPC) and forward looking expectations with the lagged inflation rate to analyse the relative importance of the backward and forward looking expectations in the US inflation dynamics. With this hybrid NKPC, Gali and Gertler (1999) found that although the coefficients of backward looking expectations are significant, their effects are relatively smaller compared to forward looking expectations. This result is interpreted by them as an indirect validation of the micro-optimisation theory behind the NKPC. They also found that the share of wages is a better proxy for marginal cost than the output gap.

However, Rudd and Whelan (2006, 2007) questioned these findings. They showed that when a correctly specified NKPC, with model consistent rational expectations, is estimated the coefficients of forward looking expectations are insignificant and inflation is highly persistent. They did not find much difference in the proxies used for marginal costs. Rudd and Whelan's findings have important policy implications because if inflation is highly persistent, that is backward looking expectations dominate inflation dynamics, then, the effects of nominal shocks on the real variables will also persist and anti-inflation policies are costly; see, for example, Guerron-Quintana (2011). Therefore, it is important to examine with alternative procedures if backward or forward looking expectations dominate the dynamics of inflation.

However, a neglected issue in this debate concerns the time series properties of its key variables viz., the rate of inflation, variables used to proxy marginal cost and any survey measures, if used, to proxy the expected rate of inflation. Recently Boug, Cappelen and Swensen (2010) and Rao and Paradiso (2011) have examined the time series properties of these variables with the US data but with different sample periods and found that they are nonstationary in their levels and stationary in first differences. Therefore, they have estimated the US NKPC with alternative time series methods of cointegration and error correction method.

The justification for the present paper is as follows. There is more than one alternative test, with various options, to determine the order of the variables and each test claims that it is

more efficient with better finite sample properties. For example, there are more than 100 alternative ways of testing for unit roots in some popular softwares like the EViews. Therefore, different conclusions are possible with different tests, options and samples.¹ For example, in the US data used by Gali and Gertler (1999) for the period 1960Q1 to 1997Q4, the standard unit root tests show that inflation rate has a unit root but the share of wages is stationary. Pesaran and Shin's (1999) ARDL method is popular with many applied workers to estimate such relationships when the order of the variables is different. However, the ARDL approach has some limitations in that the computed test statistics for cointegration may fall into a substantial inconclusive range. Furthermore, the test statics are given for sample sizes of 500 and above and their finite sample properties are not well known.

An alternative to the ARDL, and also other time series methods with all I(1) level variables, is the general to specific method (GETS) of estimating the long and short run relationships. GETS was originally developed in the 1960s at the London School of Economics (LSE) and David Hendry is its most ardent exponent and supporter.² GETS takes the view that dynamics is an empirical issue because economic theory is mostly silent on the dynamics since theory is mainly concerned with establishing equilibrium relationships between the levels of the variables. Therefore, dynamics should be estimated in a way consistent with the underlying data generation process of the variables. GETS formulations enable estimation of both the long run equilibrium and short run dynamic adjustment parameters in one step. The theory behind the relationship is used to specify the long run part of the specification in the levels of the variables and lagged changes in the variables are used to specify the short run dynamics. Some time series econometricians, especially from North America, were critical of GETS specifications because the order of the variables is different in that level variables are generally I(1) and their differences are I(0), Hendry repeatedly pointed out that such criticisms are incorrect because if the underlying economic theories are valid for the specification of the long run relationships, then, the combination of the level variables should also be stationary. Therefore, GETS specifications consist of only I(0) variables and they can be estimated with the standard classical methods.³

¹ It is not uncommon to find that many applied works devote proportionately a large amount of space to present and discuss unit root test results with several options.

² It existed as a oral tradition in the LSE undergraduate applied courses of the early 1960s before Sargan, Mizon and Hendry gave more formal econometric foundations in the mid 1960s.

³ On the use and advantages of GETS see Rao (2007) and Rao, Singh and Kumar (2010).

The implication of the above discussion to estimate the NKPC is follows. According to GETS the NKPC can be estimated with a classical method, such as the GMM, as long as the underlying micro theory on how firms set optimal product prices is valid. Therefore, this paper illustrates the use of GETS to estimate the US NKPC. Our sample is for 1978Q1 to 2010Q2 and this choice is due to some constraints on the availability of data to proxy forward looking expectations.

The rest of this paper is as follows. Section 2 examines specification issues and empirical results are presented and discussed in Section 3. Section 4 concludes.

2. Specification

For the long run Phillips curve we shall use the following Gali and Gertler's (1999) specification, based on the optimisation model with forward expectations.

$$(1) \quad \Delta \ln P = \alpha + \gamma \ln S + \beta E_t(\Delta \ln P_{t+1})$$

$$\gamma > 0; \beta = 1.$$

where $\Delta \ln P$ = rate of inflation and S = share of wages to proxy marginal cost. An alternative proxy for marginal cost is the well known output gap (GAP), measured as the difference between the logs of actual and potential outputs. GETS formulation assumes that the observed change in the dependent variable, in our case the rate of acceleration of inflation ($\Delta^2 \ln P$), is due to two reasons. Firstly, if in the previous period the actual rate of inflation did not fully adjust to its equilibrium rate in equation (1), inflation rate in the current period changes to close partly this gap. Secondly, inflation rate may also change due to changes in its determinants viz., wage share and the expected rate of inflation. Therefore, the GETS specification of the NKPC, augmented with lags of the dependent variable to capture persistence, is:

$$(2) \quad \Delta^2 \ln P_t = -\lambda [\Delta \ln P_{t-1} - (\gamma \ln S_{t-1} + \beta E_{t-1}(\Delta \ln P_t))] + \sum_{i=1}^{n1} \mu_{1i} \Delta^2 \ln P_{t-i} + \sum_{j=1}^{n2} \mu_{2j} \Delta S_{t-j} + \sum_{m=1}^{n3} \mu_{3-m} \Delta E_{t-j}(\Delta \ln P_{t-j}) + \varepsilon_t$$

$$\varepsilon \sim N(0, \sigma^2)$$

The Gali and Gertler method of proxying forward looking expectations with the actual value of the rate of inflation would cause serious estimation problems because of the presence of both the current and lagged inflation rates on the right hand and especially when β is expected to be unity. The matrix of the coefficients will be singular and estimation breaks down. Therefore, it is necessary to proxy the expected rate of inflation with some survey based measure of the expected rate of inflation. Although in the USA there are four survey based estimates of the expected rate of inflation, we selected the survey data of the University of Michigan from 1978Q1 to 2010Q2 for two reasons. Firstly, Baghestani and Noori (1988) have shown that these are consistent with Pearce's (1978) criteria of rational expectations. Secondly, consistent data without major revisions to the survey methods, are available from 1978Q1 for the Michigan survey. Denoting this as *MICH*, equation (2) can be specified as:

$$(3) \quad \begin{aligned} \Delta^2 \ln P_t = & -\lambda [\Delta \ln P_{t-1} - (\gamma \ln S_{t-1} + \beta MICH_{t-1})] \\ & + \sum_{i=1}^{n1} \mu_{1i} \Delta^2 \ln P_{t-i} + \sum_{j=1}^{n2} \mu_{2j} \Delta S_{t-j} + \sum_{m=1}^{n3} \mu_{3m} \Delta MICH_{t-m} + \varepsilon_t \\ & \varepsilon \sim N(0, \sigma^2) \end{aligned}$$

where *MICH* = University of Michigan's forecast of the median rate of expected inflation four quarters ahead. Our inflation rate is measured with the core CPI and also with the standard GDP deflator. Further details of the definitions of the variables and sources of data are in the appendix.

3. Empirical Results

Equation (3) is estimated for the period 1978Q1 to 2010Q2 with GMM and with similar instruments used by Gali and Gertler (1999). The results, with inflation measured with the core CPI, are in Table 1. For convenience this table is split into two pages. In columns (1) and (2) the driving force of inflation is proxied with the share of wages ($\ln S$). In column (1) estimates with the lagged changes of the three variables is shown. In column (2) estimates the instrument are reduced by dropping the spread between the short and long run interest rates (*SPREAD*), to see how sensitive are the estimates to the choice of instruments. The estimates of the coefficients in column (2) are similar to those in column (1). We shall discuss these estimates shortly. Some alternative proxies for the driving force are used and their estimates

are in columns (3) to (10). These proxies are four measures of the output gap ($GAP1$, $GAP2$, $GAP3$ and GAP) and the log of the probability of finding a job by newly unemployed workers ($\ln JFP$).

Output gap is the difference between the logs of actual and potential GDP and potential GDP is computed in four alternative ways viz., as a liner trend ($GAP1$), as a quadratic trend ($GAP2$), with the univariate unobserved component model ($GAP3$) and as HP filtered (GAP). However, results with $GAP1$ and $GAP2$ are unsatisfactory. The coefficient of $GAP1$ turned out to be negative and of $GAP2$, although positive, is insignificant. To conserve space only results with $GAP3$ and GAP are reported in Table 1. The unsatisfactory results with $GAP1$ and $GAP2$, based on deterministic trends, may be due to shifts in the trend and they may have been adequately captured by GAP and $GAP3$, with the stochastic trends. Furthermore, a few alternative instrumental variables (shown in the notes to Table 1) are used to check the sensitivity of the estimates.

In columns (1) to (8), where the four specifications are estimated with two alternative sets of instrumental variables, the adjustment coefficient λ has the expected negative sign and significant in all these estimates. Its absolute values ranged from 0.25 to 0.11 depending on the selected driving force and instruments. However, its estimate is more sensitive to the selected driving force than the instruments. The speeds of adjustment with both the wage share and job finding probability are similar and faster than with estimates with the two gaps. Estimates of the coefficient of the expected rate of inflation (β) are closer in all estimates. Wald tests show that β is not significantly different from the expected value of unity, at the 5% or 1% levels, with $\ln S$, GAP and $GAP3$, but exceeds unity in columns (7) and (8) with $\ln JFP$. Therefore, these two equations are reestimated with the restriction that $\beta = 1$ and these estimates are in columns (9) and (10). In these constrained estimates there are only small changes. While estimates of λ have decreased, those of persistence (μ_1) have increased.

Table 1 GMM Estimates

US NKPC 1978Q1-2010Q2 for Core Inflation

(Notes and Columns (7) to (10) are in the next page)

$$\Delta^2 \ln P_t = -\lambda [\Delta \ln P_{t-1} - (\gamma X_{t-1} + \beta MICH_{t-1})] + \sum_{i=1}^{n1} \mu_{1i} \Delta^2 \ln P_{t-i} + \sum_{j=1}^{n2} \mu_{2j} \Delta X_{t-j} + \sum_{m=1}^{n3} \mu_{3m} \Delta MICH_{t-m}$$

	(1) (X = ln S)	(2) (X = ln S)	(3) (X = GAP)	(4) (X = GAP)	(5) (X = GAP3)	(6) (X = GAP3)
<i>Interc.</i>	-25.4568 (0.832)***	-24.9934 (0.876)***	-0.1283 (0.101)	-0.1778 (0.103)*	-0.1230 (0.102)	-0.1789 (0.104)*
λ	-0.2402 (0.092)***	-0.2436 (0.092)***	-0.1140 (0.068)*	-0.1481 (0.082)*	-0.1195 (0.070)*	-0.1536 (0.082)*
γ	0.2276 (0.064)***	0.2199 (0.063)***	0.5270 (0.242)**	0.4439 (0.198)**	0.4916 (0.200)**	0.4166 (0.198)***
β	1.1192 (0.091)***	1.1471 (0.086)***	1.2649 (0.250)***	1.2993 (0.147)***	1.2519 (0.235)***	1.2935 (0.139)***
μ_{11}	0.3041 (0.116)***	0.3335 (0.121)***	0.1926 (0.133)	0.2064 (0.110)	0.1774 (0.126)	0.1925 (0.104)*
μ_{21}	0.0248 (0.057)	0.0009 (0.060)	0.0021 (0.076)	0.0082 (0.063)	0.0136 (0.068)	0.0193 (0.056)
μ_{31}	-0.0734 (0.076)	-0.0822 (0.072)	-0.0088 (0.055)	-0.0558 (0.067)	-0.0180 (0.057)	-0.0661 (0.066)
<i>J</i>	0.423	0.331	0.294	0.354	0.245	0.375
<i>Wald Test</i> $H_0 : \beta = 1$	0.193	0.089	0.292	0.043	0.286	0.037
$\overline{R^2}$	0.9755	0.9754	0.9744	0.9746	0.9748	0.9750
IVs	IV(1)	IV(2)	IV(1)	IV(2)	IV(1)	IV(2)

Table 1 GMM Estimates (Continued)
US NKPC 1978Q1-2010Q2 for Core Inflation

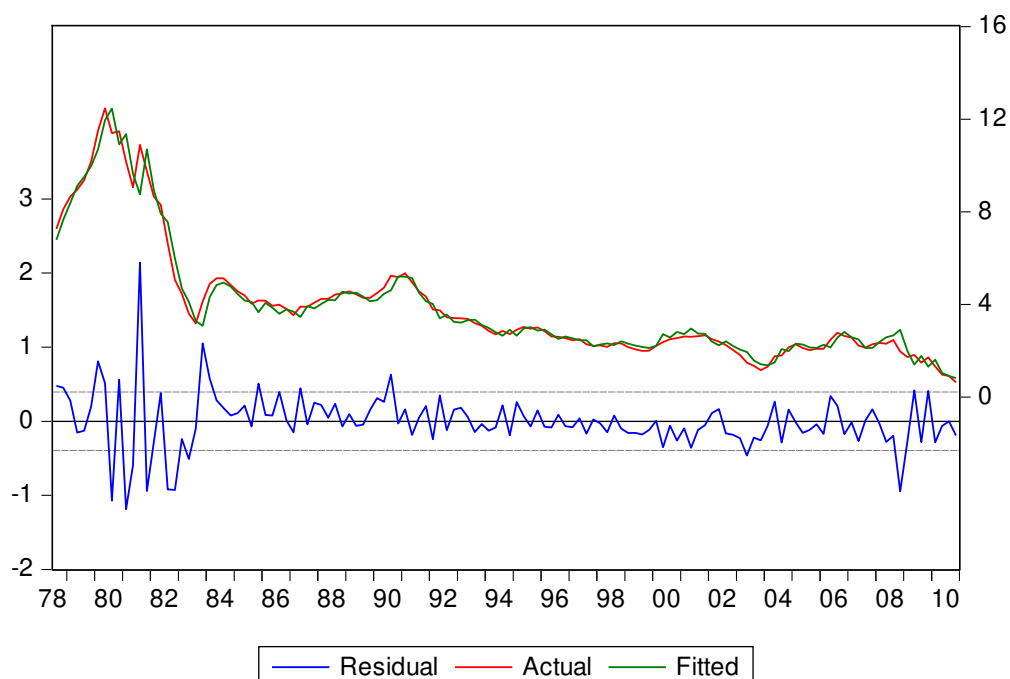
$$\Delta^2 \ln P_t = -\lambda [\Delta \ln P_{t-1} - (\gamma X_{t-1} + \beta MICH_{t-1})] + \sum_{i=1}^{n1} \mu_{1i} \Delta^2 \ln P_{t-i} + \sum_{j=1}^{n2} \mu_{2j} \Delta X_{t-j} + \sum_{m=1}^{n3} \mu_{3m} \Delta MICH_{t-m}$$

	(7) (X = ln JFP)	(8) (X = ln JFP)	(9) (X = ln JFP)	(10) (X = ln JFP)
<i>Interc.</i>	0.7302 (0.363)**	0.6715 (0.325)**	0.7592 (0.226)***	0.6882 (0.251)***
λ	-0.2481 (0.081)***	-0.2264 (0.075)***	-0.1921 (0.046)***	-0.1740 (0.054)***
γ	3.8200 (0.410)***	3.8814 (0.334)***	4.2186 (0.824)***	4.2007 (0.736)***
β	1.2102 (0.052)***	1.2202 (0.035)***	1	1
μ_{11}	0.1373 (0.045)***	0.1860 (0.045)***	0.1740 (0.127)	0.2429 (0.119)**
μ_{21}	0.1827 (0.528)	0.2970 (0.443)	0.3285 (0.543)	0.3754 (0.623)
μ_{31}	-0.1521 (0.087)	-0.1203 (0.076)	-0.0857 (0.069)	-0.0530 (0.075)
<i>J</i>	0.301	0.315	0.357	0.324
<i>Wald Test</i> $H_0 : \beta = 1$	0.00	0.00	<i>constrained</i>	<i>constrained</i>
\bar{R}^2	0.9774	0.9774	0.9764	0.9763
IVs	IV(3)	IV(4)	IV(3)	IV(4)

Notes: IV(1)= ln S_{t-2} , $MICH_{t-2}$, $\Delta \ln P_{t-2}$, $\Delta^2 \ln P_{t-1}$, $\Delta^2 \ln PPI_{t-2}$, $\Delta \ln S_{t-1}$, $\Delta MICH_{t-1}$, $SPREAD_{t-2}$. IV(2) = same as IV(1) without $SPREAD_{t-2}$. IV(3) = same as IV(1) but with $GAP3_{t-2}$ instead of $SPREAD_{t-2}$. IV(4) = same as IV(3) without $GAP3_{t-2}$. Adjusted R-bar square is for the constrained version of the equation where the dependent variable is the rate of inflation. $GAP1$ has the wrong sign and $GAP2$ is not statistically significant and for these reasons we do not report these results.

In the dynamics part only the coefficients of the lagged dependent variable is significant in all the estimates in columns (1) to (10), thus supporting our modification that the effects of persistence are transitory and can be captured with the lagged dependent variable. The coefficients of the lagged changes of the other two explanatory variables are insignificant in all estimates. Deleting these insignificant variables did not result in any significant changes to the estimates of the other parameters and are not reported to conserve space. Alternative estimates with different sets of instrumental variables did not have any significant effects and the Hansen J test shows that the instruments are over-identified. Furthermore, all our instruments are lagged values and are more likely to be independent of the error terms of the equations. The pseudo R -bar squares are computed by re-estimating all the equations by making the rate of inflation, instead of its change, as the dependent variable. These are very high and Figure 1 shows a close relationship between the actual and predicted rates of inflation with the equation in column (1). Plots with other estimates are similar and not shown to conserve space. The minor over prediction of inflation around the late 2007 and early 2008 is perhaps due to the financial crisis and its deflationary effects.

Figure 1
Actual and Predicted Inflation



On the basis of these results we may conclude: (a) the optimisation theory underlying the NKPC, with forward looking expectations, is supported by our estimates with alternative variables to proxy the driving force of inflation and marginal cost. In other empirical studies one or another of these measures had a wrong or insignificant coefficient. However, it should be noted that our estimates with gap measures with deterministic trends are also found to be unsatisfactory; (b) the expectations hypothesis is validated by our results and (c) our assumption that persistence effects are transitory is supported by the results in that the coefficients of the lagged dependent variable are all significant. However, these persistence effects are not as high as in other empirical studies and their estimates are sensitive to the selected proxy for the driving force.

These above findings are further supported by the results in Table 2 where inflation is measured with the GDP deflator. The coefficient of GAP2, where potential output is measured with the quadratic time trend, is significant now but the Wald test rejected the null that $\beta = 1$. Estimates of the adjustment coefficients are all significant and closer than their estimates in Table 1. They imply that about 15% to 17% adjustment in inflation towards its equilibrium rate takes place in one quarter.

Estimates of β and the persistence coefficient μ_{11} , with alternative measures of the driving force, are also much closer in Table 2 than in Table 1. The null that $\beta = 1$ is not rejected by the Wald test in all but with GAP2 in column (4). Estimates of the persistence coefficients ranged from about 0.28 (column (4)) to 0.37 (column (2)). All the estimates of the coefficients are less sensitive to the selected instruments and the instruments are over-identified. All in all these results support the three conclusions drawn earlier.

Table 2 GMM Estimates

US NKPC 1978Q1-2010Q2 for Inflation with GDP Deflator

$$\Delta^2 \ln P_t = -\lambda [\Delta \ln P_{t-1} - (\gamma X_{t-1} + \beta MICH_{t-1})] + \sum_{i=1}^{n1} \mu_i \Delta^2 \ln P_{t-i} + \sum_{j=1}^{n2} \mu_{2j} \Delta X_{t-j} + \sum_{m=1}^{n3} \mu_{3m} \Delta MICH_{t-m}$$

 $\Delta \ln P =$ Inflation with GDP deflator

	(1) ($X = \ln S$)	(2) ($X = \ln S$)	(3) ($X = GAP$)	(4) ($X = GAP2$)	(5) ($X = GAP3$)	(6) ($X = \ln JFP$)
<i>Interc.</i>	-10.6342 (0.630)*	-11.0137 (0.619)*	-0.1556 (0.057)***	-0.1967 (0.054)***	-0.1619 (0.055)***	0.1362 (0.146)
λ	-0.1731 (0.039)***	-0.1725 (0.038)***	-0.1567 (0.030)***	-0.1701 (0.028)***	-0.1658 (0.027)***	-0.1709 (0.032)***
γ	0.1309 (0.075)*	0.1362 (0.071)*	0.2118 (0.177)	0.1184 (0.028)***	0.1833 (0.116)*	1.6970 (0.768)**
β	1.0270 (0.112)***	1.0182 (0.111)***	1.1144 (0.067)***	1.1737 (0.062)***	1.1161 (0.059)***	1.0757 (0.062)***
μ_{11}	0.3633 (0.103)***	0.3677 (0.101)***	0.3216 (0.056)***	0.2777 (0.051)***	0.3101 (0.052)***	0.3007 (0.081)***
μ_{21}	0.0117 (0.035)	0.0087 (0.033)	0.1134 (0.027)***	0.1189 (0.031)***	0.1232 (0.026)***	0.5584 (0.321)*
μ_{31}	-0.0561 (0.069)	-0.0572 (0.071)	-0.1150 (0.038)***	-0.1233 (0.041)***	-0.1243 (0.039)***	-0.0780 (0.047)*
<i>J</i>	0.823	0.782	0.364	0.406	0.361	0.783
<i>Wald Test</i> $H_0: \beta = 1$	0.809	0.870	0.092	0.006	0.053	0.223
\bar{R}^2	0.9831	0.9831	0.9841	0.9849	0.9845	0.9833
IVs	IV(1)	IV(2)	IV(2)	IV(2)	IV(2)	IV(2)

Notes: IV(1)= $\ln S_{t-2}, MICH_{t-2}, \Delta \ln P_{t-2}, \Delta^2 \ln P_{t-1}, \Delta^2 \ln PPI_{t-1}, \Delta \ln S_{t-1}, \Delta MICH_{t-1}, GAP3_{t-2}$. IV(2) = same as IV(1) without $GAP3_{t-2}$. $GAP1$ has the wrong sign and for this reason we do not report the results.

4. Conclusions

In this paper we have used the LSE and Hendry GETS approach to specify and estimate the US NKPC. Some new features of our approach, among which the most important ones are a distinction between the long run theory based expectations augmented Phillips curve and the introduction of persistence effects into the dynamics of this relationship. Our major findings are that alternative proxies for the inflation driving force are all satisfactory with the exception of gaps estimated with the deterministic trends; the theory based NKPC with the forward looking expectations is sound and our results support this; the long run Phillips curve is vertical and therefore the real effects of nominal shocks are significant only during the transition of the economy between its equilibrium states and the duration of this transition period seems to be shorter than estimates with arbitrarily adding the lagged inflation rate to the NKPC. However, more reliable estimates of the duration of this transition period would be useful and can be estimated by simulating the effects of nominal shocks with an aggregate demand and supply model in which the NKPC plays an important role. Therefore, our conclusion on the duration of the transition period should be treated with caution.

Data Appendix
Definitions and Data Source: 1978Q1 – 2010Q2

Variable	Definition	Source
$\Delta \ln P$	Measured as $\ln \left[\frac{p_t}{p_{t-4}} \right]$ using core CPI. Consumer Price Index (All Items Less Food and Energy), Index 1982-1984=100. In Table 2 we measure inflation using GDP deflator (Index 2005=100).	research.stlouisfed.org/fred2/categories/9 research.stlouisfed.org/fred2/categories/21
$\Delta \ln PPI$	Measured as $\ln \left[\frac{ppi_t}{ppi_{t-4}} \right]$ using Producer Price Index: Finished goods (Index 1982 = 100).	research.stlouisfed.org/fred2/categories/31
$GAP, GAP 1, GAP 2, GAP 3$	GAP = output gap (Nonfarm business sector, index 1992 = 100) using the Hodrick-Prescott filter with a smoothing parameter of 1600. $GAP 1$ = output gap (Nonfarm business sector, index 1992 = 100) using a linear trend. $GAP 2$ = output gap (Nonfarm business sector, index 1992 = 100) using a quadratic trend. $GAP 3$ = output gap (Nonfarm business sector, index 1992 = 100) using univariate unobserved component models technique.	research.stlouisfed.org/fred2/categories/2
$\ln S$	Labour's Share of Income (Nonfarm Business Sector, Index 2005=100) expressed in natural log and multiplied for 100.	www.bls.gov/data
$\ln JFP$	Log of Job Finding Probability. Constructed from the number of unemployed workers (U_t), the number of short term (1 month, U_{t+1}^S) unemployed workers and the number of unemployed workers next month (U_{t+1}) according to Shimer (2005): $JFP = 1 - \frac{U_{t+1} - U_{t+1}^S}{U_t}$	www.bls.gov
$SPREAD$	Difference between 10-Year Treasury constant maturity rate and Federal Fed Funds.	research.stlouisfed.org/fred2/categories/22
$MICH_t = E_t(\Delta \ln P_{t+1})$	Median expected price change next 12 months, Survey of Consumers.	www.sca.isr.umich.edu

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Table 1 GMM Estimates

US NKPC 1978Q1-2010Q2 for Core Inflation

$$\Delta^2 \ln P_t = -\lambda [\Delta \ln P_{t-1} - (\gamma X_{t-1} + \beta MICH_{t-1})] + \sum_{i=1}^{n1} \mu_{1i} \Delta^2 \ln P_{t-i} + \sum_{j=1}^{n2} \mu_{2j} \Delta X_{t-j} + \sum_{m=1}^{n3} \mu_{3m} \Delta MICH_{t-m}$$

	(1) (X = ln	(2) (X = ln	(3) (X = G	(4) (X = G	(5) (X = G	(6) (X = G	(7) (X = ln	(8) (X = ln	(9) (X = ln	(10) (X = ln
<i>Interc</i>	- 25.45 68 (0.832))***	- 24.99 34 (0.876))***	- 0.1283 (0.101)	- 0.1778 (0.103) *	-0.1230 (0.102)	-0.1789 (0.104) *	0.7302 (0.363)* *	0.6715 (0.325)* *	0.7592 (0.226)* **	0.6882 (0.251)* **
λ	- 0.240 2 (0.092))***	- 0.243 6 (0.092))***	- 0.1140 (0.068) *	- 0.1481 (0.082) *	-0.1195 (0.070) *	-0.1536 (0.082) *	-0.2481 (0.081)* **	-0.2264 (0.075)* **	-0.1921 (0.046)* **	-0.1740 (0.054)* **
γ	0.227 6 (0.064))***	0.219 9 (0.063))***	0.5270 (0.242) **	0.4439 (0.198) **	0.4916 (0.200) **	0.4166 (0.198) ***	3.8200 (0.410)* **	3.8814 (0.334)* **	4.2186 (0.824)* **	4.2007 (0.736)* **
β	1.119 2 (0.091))***	1.147 1 (0.086))***	1.2649 (0.250) ***	1.2993 (0.147) ***	1.2519 (0.235) ***	1.2935 (0.139) ***	1.2102 (0.052)* **	1.2202 (0.035)* **	1	1
μ_{11}	0.304 1 (0.116))***	0.333 5 (0.121))***	0.1926 (0.133)	0.2064 (0.110)	0.1774 (0.126)	0.1925 (0.104) *	0.1373 (0.045)* **	0.1860 (0.045)* **	0.1740 (0.127)	0.2429 (0.119)* *
μ_{21}	0.024 8 (0.057))	0.000 9 (0.060))	0.0021 (0.076)	0.0082 (0.063)	0.0136 (0.068)	0.0193 (0.056)	0.1827 (0.528)	0.2970 (0.443)	0.3285 (0.543)	0.3754 (0.623)
μ_{31}	- 0.073 4 (0.076))	- 0.082 2 (0.072))	- 0.0088 (0.055)	- 0.0558 (0.067)	-0.0180 (0.057)	-0.0661 (0.066)	-0.1521 (0.087)*	-0.1203 (0.076)	-0.0857 (0.069)	-0.0530 (0.075)
<i>J</i>	0.423	0.331	0.294	0.354	0.245	0.375	0.301	0.315	0.357	0.324
<i>Wald Test</i> $H_0: \beta$	0.193	0.089	0.292	0.043	0.286	0.037	0.00	0.00	-	-

$\overline{R^2}$	0.975 5	0.975 4	0.9744	0.9746	0.9748	0.9750	0.9774	0.9774	0.9764	0.9763
IVs	IV(1)	IV(2)	IV(1)	IV(2)	IV(1)	IV(2)	IV(3)	IV(4)	IV(3)	IV(4)
<p>Notes: IV(1)= $\ln S_{t-2}, MICH_{t-2}, \Delta \ln P_{t-2}, \Delta^2 \ln P_{t-1}, \Delta^2 \ln PPI_{t-2}, \Delta \ln S_{t-1}, \Delta MICH_{t-1}, SPREAD_{t-2}$. IV(2) = same as IV(1) without $SPREAD_{t-2}$. IV(3) = same as IV(1) but with $GAP3_{t-2}$ instead of $SPREAD_{t-2}$. IV(4) = same as IV(3) without $GAP3_{t-2}$. Adjusted R-bar square is for the constrained version of the equation where the dependent variable is the rate of inflation. $GAPI$ has the wrong sign and $GAP2$ is not statistically significant and for these reasons we do not report these results.</p>										

Table 2 GMM Estimates

US NKPC 1978Q1-2010Q2 for Inflation with GDP Deflator

$$\Delta^2 \ln P_t = -\lambda [\Delta \ln P_{t-1} - (\gamma X_{t-1} + \beta MICH_{t-1})] + \sum_{i=1}^{n1} \mu_{1i} \Delta^2 \ln P_{t-i} + \sum_{j=1}^{n2} \mu_{2j} \Delta X_{t-j} + \sum_{m=1}^{n3} \mu_{3m} \Delta MICH_{t-m}$$

 $\Delta \ln P$ = Inflation with GDP deflator

	(1) ($X = \ln S$)	(2) ($X = \ln S$)	(3) ($X = GAP$)	(4) ($X = GAP2$)	(5) ($X = GAP3$)	(6) ($X = \ln JFP$)
<i>Interc.</i>	-10.6342 (0.630)*	-11.0137 (0.619)*	-0.1556 (0.057)***	-0.1967 (0.054)***	-0.1619 (0.055)***	0.1362 (0.146)
λ	-0.1731 (0.039)***	-0.1725 (0.038)***	-0.1567 (0.030)***	-0.1701 (0.028)***	-0.1658 (0.027)***	-0.1709 (0.032)***
γ	0.1309 (0.075)*	0.1362 (0.071)*	0.2118 (0.177)	0.1184 (0.028)***	0.1833 (0.116)*	1.6970 (0.768)**
β	1.0270 (0.112)***	1.0182 (0.111)***	1.1144 (0.067)***	1.1737 (0.062)***	1.1161 (0.059)***	1.0757 (0.062)***
μ_{11}	0.3633 (0.103)***	0.3677 (0.101)***	0.3216 (0.056)***	0.2777 (0.051)***	0.3101 (0.052)***	0.3007 (0.081)***
μ_{21}	0.0117 (0.035)	0.0087 (0.033)	0.1134 (0.027)***	0.1189 (0.031)***	0.1232 (0.026)***	0.5584 (0.321)*
μ_{31}	-0.0561 (0.069)	-0.0572 (0.071)	-0.1150 (0.038)***	-0.1233 (0.041)***	-0.1243 (0.039)***	-0.0780 (0.047)*
<i>J</i>	0.823	0.782	0.364	0.406	0.361	0.783
<i>Wald Test</i> $H_0 : \beta = 1$	0.809	0.870	0.092	0.006	0.053	0.223
\bar{R}^2	0.9831	0.9831	0.9841	0.9849	0.9845	0.9833
IVs	IV(1)	IV(2)	IV(2)	IV(2)	IV(2)	IV(2)

Notes: IV(1)= $\ln S_{t-2}, MICH_{t-2}, \Delta \ln P_{t-2}, \Delta^2 \ln P_{t-1}, \Delta^2 \ln PPI_{t-1}, \Delta \ln S_{t-1}, \Delta MICH_{t-1}, GAP3_{t-2}$. IV(2) = same as IV(1) without $GAP3_{t-2}$. *GAPI* has the wrong sign and for this reason we do not report the results.