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Recovering Historical Inflation Data from Postage Stamps Prices

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ABSTRACT	ARTICLE INFO
<p>For many developing countries, historical inflation figures are rarely available. We propose a simple method that aims to recover such figures of inflation using prices of postage stamps issued in earlier years. We illustrate our method for Suriname, where annual inflation rates are available for 1961 until 2015, and where fluctuations in inflation rates are prominent. We estimate the inflation rates for the sample 1873 to 1960. Our main finding is that high inflation periods usually last no longer than 2 or 3 years. An Exponential Generalized Autoregressive Conditional Heteroscedasticity (EGARCH) model for the recent sample and for the full sample with the recovered inflation rates shows the relevance of adding the recovered data.</p>	<p><i>Keywords:</i> inflation; postage stamps; price recovery; historical time series; EGARCH</p> <p><i>Article History:</i> Received: 17 October 2017; Accepted: 9 November 2017;</p>

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1. INTRODUCTION AND MOTIVATION

The World Bank collects annual inflation rates for all countries in the world. For developed countries, such data can be available for a long span of time, also because statistical bureaus for the countries exist for a long time.¹ For many developing countries, matters can be different. For example, Benin's first available inflation figure concerns 1993, whereas Ethiopia's first quote concerns 1966. There may be various causes for this lack of data, which can relate to a lack of institutions and the effects of decolonization.

For various reasons, one may want to have some impression of historical figures. One would perhaps want to know if a current high inflation period, which indicates a period with risky economic fluctuations, has occurred before and which measures were taken to reduce inflation. Alternatively, one may want to compare inflation patterns across countries to discern similarities or specific differences. In addition, one may want to compute historical data on real GDP, real wages, and purchasing power parity, which all involve price levels. Preferably, one would also want to have annual data without missing data in between.

Recovering historical price levels can be difficult because of lack of information on the prices of many goods and because good and services may have changed substantially over the years. Various recent studies present discussions of methods to recover historical data and the reliability of those historical statistics, see Allen et al. (2011), Bolt and Zanden (2014), Cendejas Bueno and Font de Villanueva (2015), Deaton and Heston (2010), Frankema and Waijenburg (2012), and Jerven (2009).

In the present paper, we present a new and very simple method that aims to reconstruct historical inflation rates. We seek to alleviate the issue of changing products over time by considering a product that has not much changed over time and for which prices are immediately available. This product concerns postage stamps. First, postage stamps have been issued in many countries for a long time. Next, the type of product and its use did not change much over the years. In addition, evidently, the price of the stamp is printed on the stamp; see for example Figure 1, where we present a few stamps for Suriname.

In the present paper, and only for illustrative purposes, we consider the historical prices of this South American country, which borders Guyana and Brazil. The World Bank can provide us with annual inflation rates starting in 1961. The first stamps in Suriname were issued in 1873, and hence we aim to retrieve annual inflation

rates since that year. We chose Suriname as the estimation sample because it is reasonably large, ranging from, 1961 to and including 2015, as there were many stamps issued per year, and also as there is substantial variation in the inflation figures over time. It also so happens that, at the time of writing this paper in 2016, inflation is again very high, and people may wonder how long high inflation periods typically last.

Figure 1. A few stamps issued in 1996 in Suriname (with exceptionally high prices).



The outline of our paper is as follows. Section 2 deals with the inflation rates data and with the stamps data, and provides some characteristics. Section 3 deals with two types of models to see if (changes in) postal stamps prices have explanatory value for inflation. One model is a simple regression model, while the other is a MIDAS regression, which fits annual inflation rates to quarterly stamps prices. Both models suggest strong predictive power of the stamps prices. Section 4 deals with the backward extrapolation and identifies a few historical periods with excessive inflation rates and their potential causes. Section 5 concludes.

Table 1. Inflation data. All items index annual average (source: World Bank).

1961	1.7	1971	0.2	1981	8.8
1962	2.1	1972	3.2	1982	7.3
1963	2.1	1973	12.9	1983	4.4
1964	4.2	1974	16.9	1984	3.7
1965	1.9	1975	8.4	1985	10.9
1966	4.7	1976	10.1	1986	18.7
1967	10.7	1977	9.7	1987	53.4
1968	0.2	1978	8.8	1988	7.3
1969	11.3	1979	14.8	1989	0.8
1970	2.6	1980	14.1	1990	21.7
1991	26	2001	38.6	2011	17.7
1992	43.7	2002	15.5	2012	5
1993	143.5	2003	23	2013	2
1994	368.5	2004	10	2014	3.3
1995	235.6	2005	9.9	2015	6.9
1996	-0.7	2006	11.3		
1997	7.1	2007	6.4		
1998	19	2008	14.7		
1999	98.8	2009	-0.2		
2000	59.4	2010	6.9		

2. THE DATA

The annual inflation rates for 1961 to 2015 for Suriname are retrieved from the World Bank. Table 1 presents the data, and Figure 2 visualizes the data. Clearly, the data show periods with high inflation, and 1994 stands out with an inflation rate of 368.5 per cent. This number associates with approximately 1.6 per cent per day. Periods with high inflation are 1993 to 1995 and 1999 to 2001, while 1987 was also an exceptional year. The International Monetary Fund provides a review of the potential causes for the high inflation rates in the nineties.

Figure 2. Inflation (in yearly percentages) in Suriname.

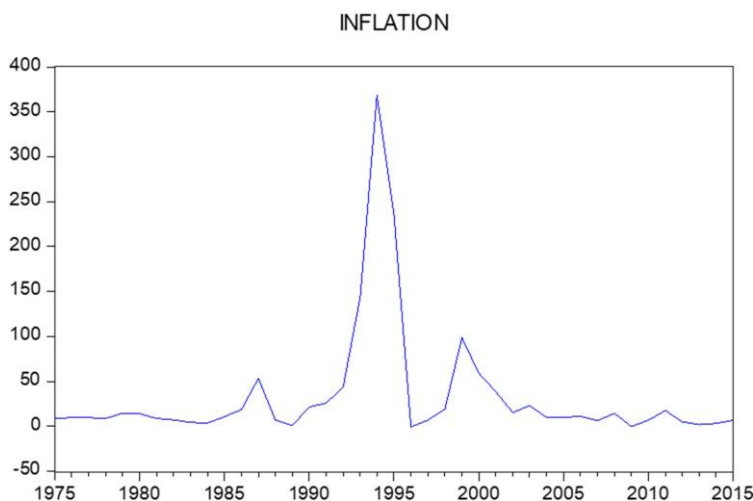


Table 2 presents the data on the postage stamps, also for the period 1961 to 2015. We consulted two catalogues, and retrieved the median prices of all the stamps in each year. The first catalogue runs until 12 November 1975, a few weeks before Suriname became an independent country on 25 November 1975. Before independence, Suriname was a colony of the Netherlands since the 17th century. The first postage stamp in Suriname was issued in 1873.

Table 2. Percentage changes in the median stamp price per year (sources: for the data until and including 12 November 1975, "Specialecatalogus 2002, Postzegels van Nederland en overzeesrijksdelen" NVPH, Amsterdam, Joh Enschede, and for the data since 25 November 1975, "Officiële postzegelcatalogus," Suriname, 31ste Editie 2016, Guernsey, Uitgeverij Zonnestraal).

1961	33.33	1971	8.7	1981	50
1962	-30	1972	20	1982	-16.67
1963	0	1973	0	1983	-50
1964	3.57	1974	0	1984	33.33
1965	3.45	1975	0	1985	10
1966	33.33	1976	16.67	1986	-9.09
1967	25	1977	42.86	1987	120
1968	0	1978	0	1988	0
1969	0	1979	-10	1989	0
1970	-8	1980	33.33	1990	0
1991	0	2001	-60	2011	-12.5
1992	40.91	2002	70	2012	28.57
1993	30.65	2003	-41.18	2013	13.89
1994	4220.99	2004	-20	2014	7.32
1995	471.43	2005	-43.75	2015	6.82
1996	-61	2006	277.78		
1997	15.38	2007	17.65		
1998	61.11	2008	50		
1999	148.28	2009	0		
2000	177.78	2010	33.33		

The data in Table 2 show that, like inflation, the percentage changes in postal stamp prices can be substantial. The observation in 1994 stands out with a value of 4220.99 per cent. In various years, we also see that price changes were equal to zero.

Figures 3 and 4 provide scatters of inflation versus the percentage changes in stamps prices. The first scatter covers all data, whereas Figure 4 concerns the data except 1994. A first impression is that there may exist a positive relationship between the two series.

Figure 3. Scatter of inflation versus percentage changes in stamps prices (all observations).

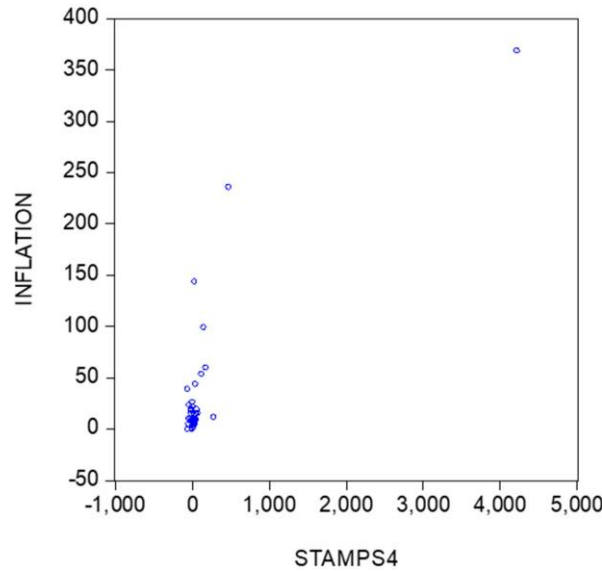
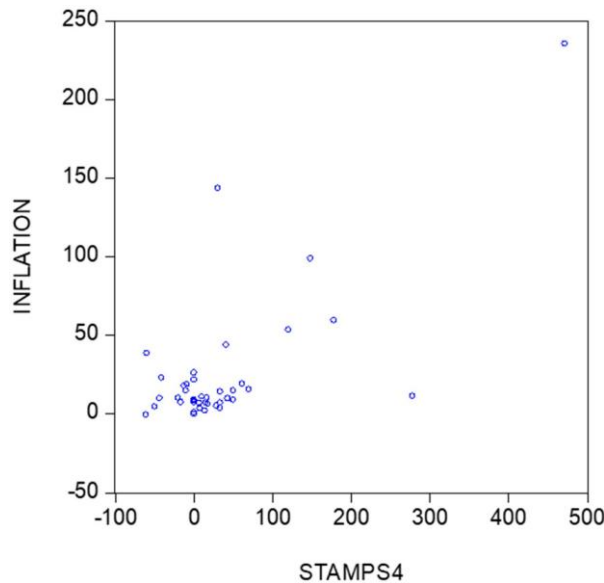


Figure 4. Scatter of inflation versus percentage changes in stamps prices (excluding 1994).



3. TWO ECONOMETRIC MODELS

In this section, we consider two econometric models to link inflation with changes in stamp prices. We denote annual inflation as Y_T and the annual changes in stamps prices as X_T . For inflation, we have only annual data, but for the stamps prices, we can also construct quarterly data. Each year, a range of stamps was issued, and this allows us to construct quarterly percentage changes, as the catalogues give the exact dates (day, month, and year) of issue. There are now two ways to compute these changes in postal stamps prices. The first refer a current quarter with the previous quarter—that is, the first-order differences. We denote these as $X_{q,T}^1$, where T again corresponds with years and q with the quarters 1, 2, 3, and 4. The data appear in Figure 5. The second type of percentage changes concern the differences between a current quarter and the same quarter the year before. We denote these as $X_{q,T}^4$, as these concern fourth-order differences. The data are depicted in Figure 6.

Figure 5. Quarter-to-quarter changes in stamps prices for MIDAS regression.

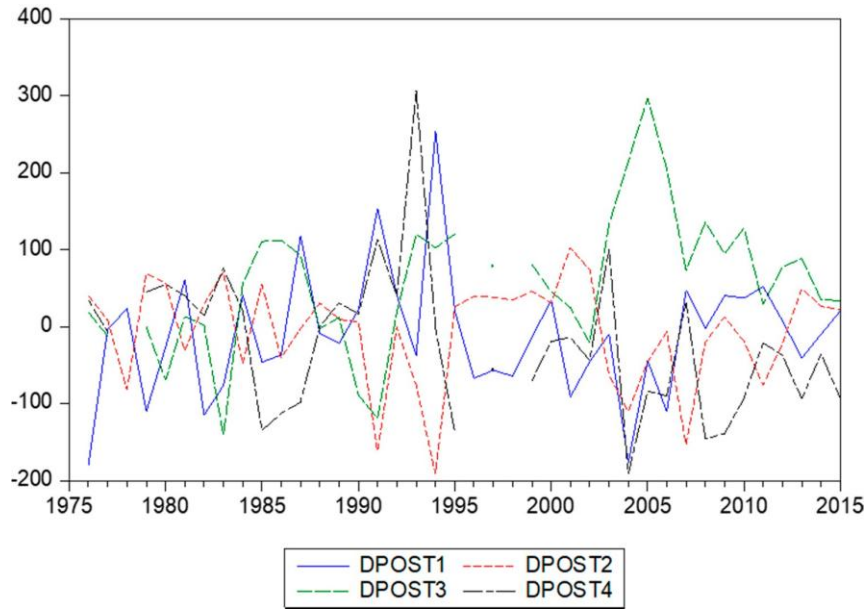
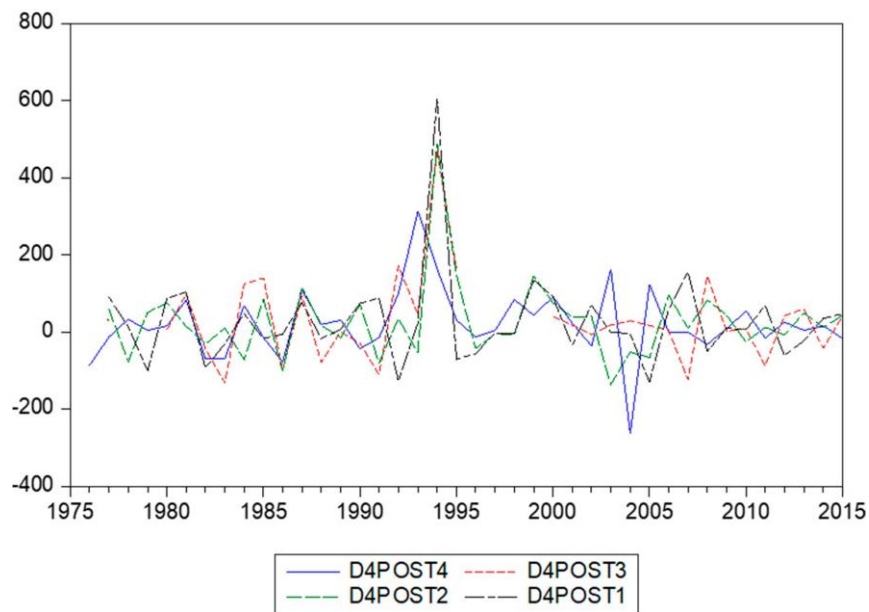


Figure 6. Annual changes in stamps prices, observed per quarter, for MIDAS regression.



At first, we allow for potential differences across models for the periods before and after the date of independence, and we thus start with the following regression model for the data from 1975 onwards, that is

$$Y_T = \mu + \alpha_0 X_T + \alpha_1 X_{T-1} + \rho Y_{T-1} + \varepsilon_T$$

where Y_T is annual inflation and as X_T are the annual changes in stamps prices. The estimation results for the sample 1975 to 2015 appear in the first panel of Table 3. The parameters are estimated using Ordinary Least Squares (OLS), with Newey-West adjusted standard errors. The tests for normality and first-order residual autocorrelation indicate that the model can be improved. A closer look at the residuals reveals that there are (at least) two very large residuals, concerning 1993 and 1999. The second panel of Table 3 displays the estimation results for the case where lagged inflation is deleted (as it was not significant) and where the observations in 1993 and 1999 are not included. The test results now suggest that the model is appropriately specified. The last panel of Table 3 presents the results for the same model, but now for the sample starting in 1961. Even though the tests do diagnose some problems with the errors (which are due to some modest outliers), the parameter estimates for current and lagged changes in stamps prices are remarkably constant across models and samples.

Table 3. Estimation results for various regression models relating inflation with lagged inflation and current and lagged percentage changes in stamps prices. Estimated Newey-West adjusted standard errors are in parentheses.

Variable	Sample	Sample	Sample
	1975–2015	1975–2015 (without 1993, 1999)	1961–2015 (without 1993, 1999)
Intercept	15.683 (5.398)	11.032 (2.796)	9.288 (2.171)
Y_{T-1}	0.026 (0.162)		
X_T	0.083 (0.005)	0.085 (0.001)	0.085 (0.001)
X_{T-1}	0.040 (0.013)	0.043 (0.001)	0.044 (0.001)
R^2	0.858	0.967	0.964
<i>p</i> value tests			
Normality	0.000	0.131	0.000
Autocorrelation	0.010	0.037	0.008

To see if data that are more detailed can lead to better models, we consider two MIDAS models. The first model is

$$Y_T = \mu + \rho Y_{T-1} + \beta_0 X_{4,T}^1 + \beta_1 X_{3,T}^1 + \beta_2 X_{2,T}^1 + \beta_3 X_{1,T}^1 + \beta_4 X_{4,T-1}^1 + \beta_5 X_{3,T-1}^1 + \varepsilon_T$$

where $X_{q,T}^1$ is the first-order differenced median stamps prices, where T again corresponds with years and q with the quarters 1, 2, 3, and 4. The OLS estimation results, with Newey-West adjusted standard errors, for this model are given in the left panel of Table 4. The diagnostic tests for normality and first-order residual autocorrelation suggest that this model is adequately specified.

Table 4. Estimation results for the (unrestricted) MIDAS regression models, sample runs from 1975 to and including 2015. Estimated Newey-West adjusted standard errors are in parentheses. $X_{q,T}$ are the first-order differenced median stamps prices, where T again corresponds with years and q with the quarters 1, 2, 3, and 4, and differenced median stamps price. $X_{q,T}^4$ are the fourth-order differenced median stamps prices.

Variable	Version 1		Variable	Version 2	
Intercept	1.828	(3.240)	Intercept	0.315	(2.811)
Y_{T-1}	0.554	(0.037)	Y_{T-1}	0.893	(0.134)
$X_{4,T}^1$	0.240	(0.066)	$X_{4,T}^4$	0.190	(0.065)
$X_{3,T}^1$	0.209	(0.060)	$X_{3,T}^4$	0.051	(0.099)
$X_{2,T}^1$	0.190	(0.085)	$X_{2,T}^4$	-0.023	(0.078)
$X_{1,T}^1$	0.402	(0.068)	$X_{1,T}^4$	0.232	(0.061)
$X_{4,T-1}^1$	0.430	(0.073)	$X_{4,T-1}^4$	0.049	(0.050)
$X_{3,T-1}^1$	0.319	(0.100)	$X_{3,T-1}^4$	0.084	(0.050)
			$X_{2,T-1}^4$	-0.264	(0.098)
R^2	0.932			0.944	
<i>p</i> value tests					
Normality	0.184			0.645	
Autocorrelation	0.149			0.740	
RMSPE, in sample	19.305			18.363	

The second MIDAS model is

$$Y_T = \mu + \rho Y_{T-1} + \beta_0 X_{4,T}^4 + \beta_1 X_{3,T}^4 + \beta_2 X_{2,T}^4 + \beta_3 X_{1,T}^4 + \beta_4 X_{4,T-1}^4 + \beta_5 X_{3,T-1}^4 + \beta_6 X_{2,T-1}^4 + \varepsilon_T$$

where $X_{q,T}^4$ are the fourth-order differenced median stamps prices, where T again corresponds with

years and q with the quarters 1, 2, 3 and 4. The OLS estimation results, with Newey-West adjusted standard errors, are in the second panel of Table 4. This MIDAS model also seems to fit the data well. When we compare the in-sample prediction errors, we see that the difference in accuracy is small. Given that the diagnostic tests indicate appropriate models, we learn that adding higher frequency terms to the regression models leads to more accuracy.

Overall, the estimation results in this section show that inflation and changes in stamps prices are strongly connected, at least here for the case of Suriname. Because the frequency of publication of the postal stamps before 1975 is lower, the MIDAS models however are not useful for the purpose we have in mind, namely recovering historical inflation rates, despite the increased accuracy they offered.

4. RECOVERY OF HISTORICAL INFLATION RATES

Given that, a suitable model for annual inflation rates appears to be

$$Y_T = \mu + \alpha_0 X_T + \alpha_1 X_{T-1} + \varepsilon_T$$

we will use the parameter estimates in the final panel of Table 3 to make backward predictions for inflation.

That is, for the sample 1873 to 1960, we compute

$$\hat{Y}_T = 9.288 + 0.085X_T + 0.044X_{T-1} \tag{1}$$

Figure 7 displays the observations on the changes in the postal stamps prices for these years,

which were collected from the first mentioned catalogue in Table 2. In addition, Figure 8 displays the estimated historical inflation observations.

Figure 7. Percentage changes in postal stamps prices, 1873–1960.

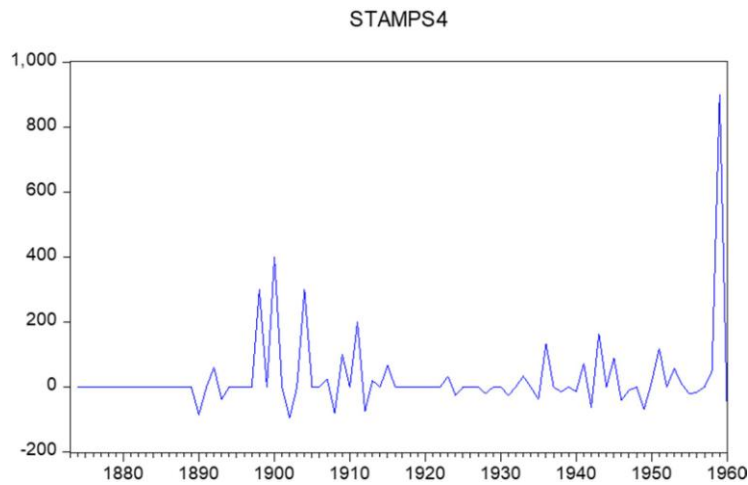
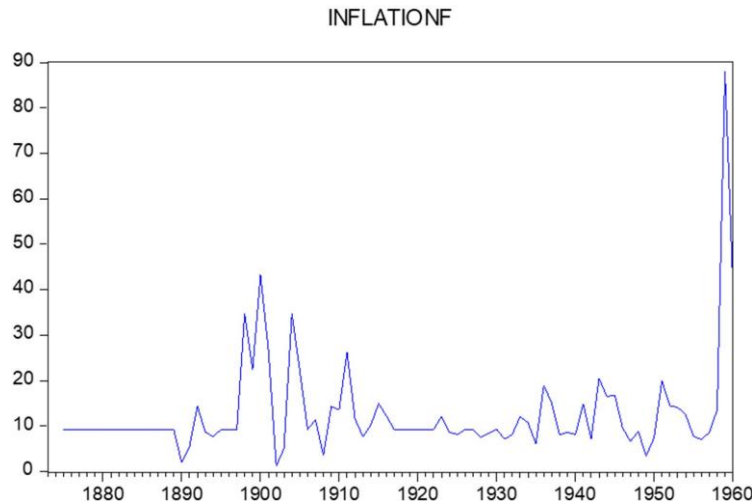


Figure 8. Predictions of inflation rates for 1873 to 1960.



We see that high inflation periods each time last for usually one or one years, and at most four years (Figure9). Specifically, if we define a high inflation period as a period with inflation higher than 10 per cent, the average length of such a period is 1.45 using the fitted sample from 1873–2015.

Figure 9. Length of periods with high inflation rates (inflation above 10 per cent) and their frequency of occurrence, based on fitted sample of 1873–2015.

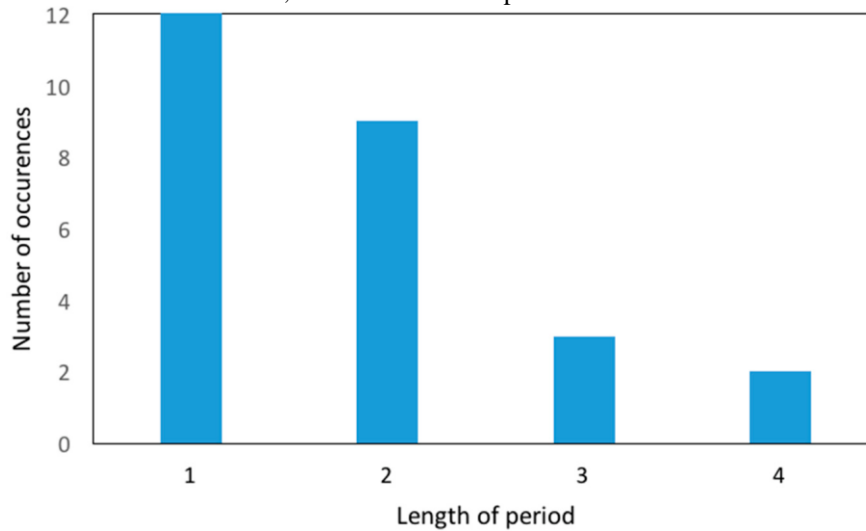
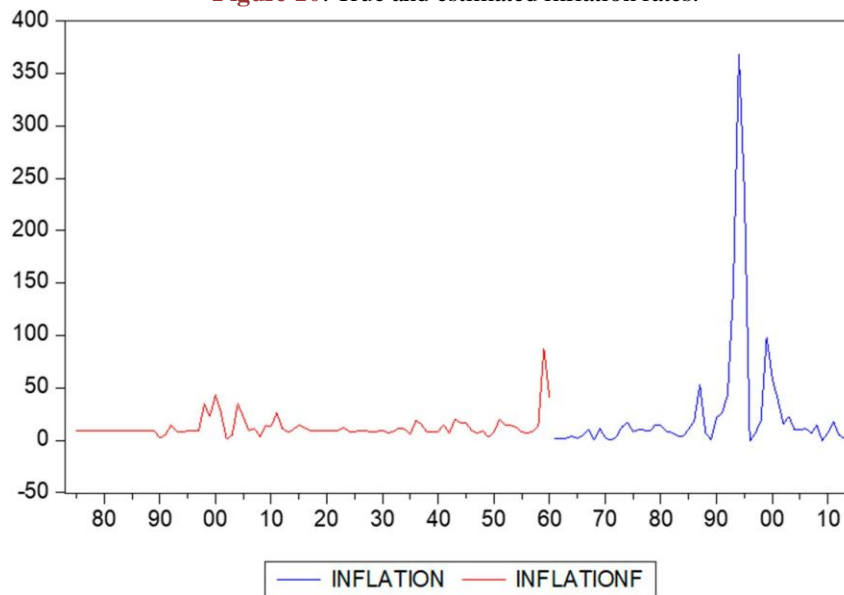


Figure10displays the actual data from 1961 onward and the estimated data back to 1873 in one single graph. Clearly, the data in the nineties are exceptional. Postal stamps experienced enormous price changes during the 1990s. This period exactly coincides with the hyperinflation that Suriname experienced during that time.

Figure 10. True and estimated inflation rates.



However, various other periods in our forecasting sample are characterized by high inflation rates. Consider for example the 1900s, 1930s, 1940s, and 1950s (specifically 1957). In Table5, more details on these periods are discussed. The high inflation rates around 1900 coincides with the gold rush at the same time at the Lawa River in Suriname, high amounted to higher tariffs for transportation (see Van Velzen and Hoogbergen(2013)). The higher inflation rates in the 1930s can be related to a period of economic decline and austerity measures, causing social upheaval (Hoefte2013). The developments in 1940s can be most likely attributed to the consequences of the Second World War. The very highinflation rates as indicated by the peak at 1957 occur in the same year as the establishment of the central bank of Suriname, but certainly coincides with the so-called Eisenhower recession.

Table 5. Estimated historical episodes with high inflation.

Years	Potential Causes
1900s	Gold rush (Lawa railway construction)
1930s	Economic decline, social upheaval in the form of riots
1940s	WW II
1957	Establishment of the Central Bank of Suriname, Brokopondo-agreement with Alcoa and Eisenhower Recession

Finally, Table6 presents the estimation results of an EGARCH model, measuring risk. The model reads as

$$\log(Inflation_t + 1) = \mu + \rho \log(Inflation_{t-1} + 1) + z_t \sqrt{h_t}$$

where $Inflation_t$ covers the full sample from 1873 to 2015, with

$$\log(h_t) = \omega + \alpha_1 |z_{t-1}| + \alpha_2 z_{t-1} + \alpha_3 \log(h_{t-1})$$

in addition, z_t is a standard normally distributed variable. We need to add 1 to where In f lationt to make sure that all observations are positive. The estimation results (with Bollerslev-Wooldridge robust standard errors in parentheses) for the sample 1961–2015 and the full sample (1873–2015) with the recovered data shows reasonably similar parameter estimates and much smaller estimated standard errors in the latter case. This seems to support the quality of the recovered inflation figures.

$$\log(Inflation_t + 1) = \mu + \rho \log(Inflation_{t-1} + 1) + z_t \sqrt{h_t}$$

with

$$\log(h_t) = \omega + \alpha_1 |z_{t-1}| + \alpha_2 z_{t-1} + \alpha_3 \log(h_{t-1})$$

Estimated using Eviews package 8.0. To reduce the effect of outliers, the data are transformed as indicated. Bollerslev-Wooldridge robust standard errors are in parentheses.

Table 6. Estimated parameters and associated standard errors in the EGARCH model.

Parameters	Sample			
	1961–2015		1873–2015	
μ	1.457	(0.200)	1.595	(0.219)
ρ	0.327	(0.089)	0.314	(0.096)
ω	−0.939	(0.249)	−0.311	(0.103)
α_1	1.073	(0.308)	0.311	(0.113)
α_2	0.711	(0.182)	0.208	(0.095)
α_3	0.659	(0.146)	0.855	(0.048)

5. CONCLUSIONS

We proposed a simple method to estimate annual historical data on inflation using changes in postal stamps prices. The method seems to work, at least for Suriname. A next step would be to reconstruct historical figures for other countries. If possible, we would want to match our estimates with actually observed inflation rates, should those data be available.

Conflicts of Interest:

The authors declare no conflict of interest.

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