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## The US real GNP is trend-stationary after all

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### ABSTRACT

This article applies the Fractional Frequency Flexible Fourier Form (FFFFF) Dickey–Fuller (DF)-type unit root test on the natural logarithm of US real GNP over the quarterly period of 1875:1–2015:2, to determine whether the same is trend- or difference-stationary. While standard and Integer Frequency Flexible Fourier Form DF-type test fails to reject the null of unit root, the relatively more powerful FFFFF DF-type test provides strong evidence of the real GNP as being trend-stationary, i.e. US output returns to a deterministic log-nonlinear trend in the long run.

### KEYWORDS

Fractional Frequency Flexible Fourier Form; structural break; unit root; US real GNP

### JEL CLASSIFICATION

C12; C22; E23

### 1. Introduction

Nelson and Plosser (1982) challenged the accepted hypothesis of output being trend-stationary, i.e. output returns to a deterministic loglinear trend in the long run. Instead, the authors provided statistical evidence that output is best viewed as difference-stationary, i.e. as a unit root process with drift. In other words, Nelson and Plosser (1982) suggested that shocks to real output have permanent rather than temporary effects. Following this piece of work by Nelson and Plosser (1982), a huge literature has emerged (for a detailed literature review, see e.g., Murray and Nelson 2000, Camacho 2011, Shelley and Wallace 2011, Cushman 2012, Hosseinkouchack and Wolters 2013, Balcilar et al. 2016 and references cited therein). This line of research applies different unit root tests to determine whether output of the US economy (measured either by real Gross Domestic Product or by real Gross National Product (GNP)) is trend-stationary or difference-stationary. More importantly, reading of the literature would indicate that it has not yet converged to a conclusive answer, with results contingent on tests and sample periods.

Against this backdrop, the objective of this study is to try and provide a definitive answer to this debate by analysing the unit root property of an unique dataset of US real GNP, covering quarterly period of 1875:1–

2015:2, which, to the best of our knowledge, is the longest possible data on US output available at quarterly frequency, i.e. the most relevant frequency at which output is measured globally.<sup>1</sup> Given that we cover over 140 years of data, we use the recently developed powerful unit root test with multiple smooth structural breaks of Omay (2015), based on a Fractional Frequency Flexible Fourier Form (FFFFF) to determine whether real GNP of the US is trend- or difference-stationary.

Recently, multiple smooth breaks have been modelled by Flexible Fourier Transforms by Becker, Enders, and Lee (2006), Enders and Lee (2012a, 2012b) and Rodrigues and Taylor (2012). The Fourier approach can capture the behaviour of a deterministic function of unknown form even if the function itself is not periodic. Hence, it works better than dummy variable methods irrespective of whether the breaks are instantaneous or smooth, and avoiding the problems of selecting the dates, number and form of breaks (Omay 2015). As Omay (2015) points out, the papers by Becker, Enders, and Lee (2006), Enders and Lee (2012a, 2012b) and Rodrigues and Taylor (2012) indicated that, due to the problem of over-filtration, single frequency component of the Fourier Transforms should be used for structural break determination. However, following Becker et al. (2004), Omay (2015) considers the Fractional Frequency version

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<sup>1</sup>Annual data of US real GDP is available from the year 1800 from the Global Financial Database.

of the Flexible Fourier Transform-based Dickey and Fuller (Augmented Dickey–Fuller (ADF), 1979)-type tests developed by Becker, Enders, and Lee (2006), Enders and Lee (2012a, 2012b) and Rodrigues and Taylor (2012), to show that the small sample properties of the proposed test are found to be better than that of the integer frequency counterpart. Given this, we compare the results from the FFFFF of the standard ADF test with the Integer Frequency Flexible Fourier Form (IFFFF) of the Dickey and Fuller (DF) test, and the basic ADF test without structural breaks, applied to the real GNP data of the US. To the best of our knowledge, this is the first such attempt based on the above-discussed three tests to try and analyse the unit root properties of the US quarterly output. The rest of the article is organized as follows: Section II presents the basics of the ADF-type unit root test with FFFF; Section III discusses the data and the empirical findings, while Section IV concludes.

## II. Unit root test with Fractional Frequency Flexible Fourier Form

The following equation is considered;

$$y_t = d(t) + \phi_1 y_{t-1} + \lambda t + \varepsilon_t \quad (1)$$

where  $\varepsilon_t$  is a stationary disturbance with variance  $\sigma^2$ , and  $d(t)$  is a deterministic function of  $t$ . We also note that the initial value of  $y_t$ , i.e.  $y_0$ , is assumed to be a fixed value, and  $\varepsilon_t$  is weakly dependent as in Enders and Lee (2012a, 2012b). As pointed out by Enders and Lee (2012a, 2012b) and Omay (2015), if the functional form of  $d(t)$  is known, it is possible to estimate Equation 1 and to test the null hypothesis of a unit root. Further, when the functional form of  $d(t)$  is unknown, any test for  $\phi_1 = 1$  is difficult if  $d(t)$  is miss-identified. The tests of Enders and Lee (2012a, 2012b) and Omay (2015) approximate  $d(t)$  by employing the Fourier expansion as follows:

$$d(t) = \alpha_0 + \alpha_k \sin\left(\frac{2\pi kt}{T}\right) + \beta_k \cos\left(\frac{2\pi kt}{T}\right) \quad (2)$$

where  $k$  indicates a particular frequency, and  $T$  is the number of observations. When there is no nonlinear trend, all values of  $\alpha_k = \beta_k = 0$ , which in turn, yield the ADF test as a special case.

As recommended in the literature, a specific single frequency often leads to a good approximation to a

model with structural change, and hence, we do the same. Recall that we use the fractional frequency instead of integer ones as stated in the introduction. For selecting the best fitting fractional single frequency, we follow the completely data-driven procedure of Davies (1987). The grid search method works as follows: we run a regression by using Equation 2, with the single frequency between the intervals  $0.1 \leq k^{fr} \leq k_{max}^{fr}$ ; we set  $k^{max} = 2$  as recommended in Enders and Lee (2012a, 2012b) and Omay (2015). However, for fractional frequencies, we select  $k = 0.1$  as increments of the selected frequencies. Finally, we obtain the specific (optimal)  $k = \tilde{k}^{fr}$  that minimizes the sum of square residuals of Equation 2. Formally, the testable regression is as follows:

$$\Delta y_t = \rho y_{t-1} + c_1 + c_2 t + c_3 \sin\left(\frac{2\pi k^{fr} t}{T}\right) + c_4 \cos\left(\frac{2\pi k^{fr} t}{T}\right) + e_t \quad (3)$$

## III. Data and empirical application

In this study, we use data on the real US GNP covering the period quarterly of 1875:1–2015:2 ( $T = 562$ ), with the start and end dates being purely driven by data availability. Our data on the nominal GNP and the GNP deflator, with the latter used to deflate the former to yield real values, are derived from two sources. First, the observations covering the period 1875:1–1946:4 are obtained from National Bureau of Economic Research, available for download at: <http://www.nber.org/data/abc/>; the actual sources are the tables of quarterly data corresponding to Appendix B of Gordon (1986). To the best of our knowledge, this is the only existing source for the pre-1947 quarterly data on US GNP and GNP deflator, with National Income and Product Account (NIPA) quarterly data series nonexistent before 1947. Second, data from 1947:1–2015:2 is sourced from the FRED database of the Federal Reserve Bank of St. Louis. Note that the dataset compiled by Gordon (1986) runs till 1983:4, with the base year of the GNP deflator being 1972. Given that nominal GNP and GNP deflator data based on the NIPA are available from 1947:1, we decided to use, for those variables, the FRED database, rather than the Gordon (1986) one, which, in any case, would have

ran only till 1983:4. The base year of the GNP deflator for the period 1875:1–1946:4 is updated from 1972 to 2009 to correspond to the base year of the GNP deflator based on the NIPA, so that the real GNP is ultimately in constant 2009 prices. For our various unit root tests, we work with the natural logarithm of the real GNP data (LGNP). The data has been plotted in Figure A1 in the Appendix.

Besides the FFFFF DF-type test of Omay (2015), Table 1 also reports the IFFFF DF-type test of Enders and Lee (2012a, 2012b), and the standard DF test. Not surprisingly, the standard DF test fails to reject the null of unit root with a constant and trend in the test-equation. The IFFF DF-test (with  $k = 2$ ) too fails to reject the null of unit root in LGNP at the conventional (5%) level of significance. The test does, however, provide weak evidence of trend-stationarity of the LGNP at the 10% level of significance.<sup>2,3</sup> Next, we turn to the FFFFF DF-type test of Omay (2015).

For detailed investigation of the optimal frequency ( $\hat{k}^{fr}$ ), we have used 0.1, 0.01 and 0.001 increments of the fractional frequencies to obtain 1.7, 1.670 and 1.667 values of the same, respectively. In all these cases, the null of unit root of the LGNP is rejected at the 5% level of significance, with the most significant test value being at  $k^{fr} = 1.667$ . Figure A1 plots the LGNP along with the estimated LGNP with  $k^{fr} = 2.0$  and  $k^{fr} = 1.667$ , while Figure A2 plots the residual from the corresponding fits of LGNP, i.e. the residual after detrending LGNP with a linear trend and Fourier intercepts under  $k^{fr} = 2.0$  and  $k^{fr} = 1.667$ . As

**Table 1.** Unit root tests on US real GNP (1875:1–2015:2).

$k$	$\tau_{DF-\tau}$	$\tau_{DF-\tau}^{fr}$			$DF_{\tau}$
	2.0	1.70	1.670	1.667	
	-3.828*	-4.322**	-4.338**	-4.339**	-3.075

Notes: The lag-length is 10 as selected by the Akaike Information Criterion with a maximum lag set at 18;  $DF_{\tau}$ ,  $\tau_{DF-\tau}$  and  $\tau_{DF-\tau}^{fr}$  corresponds to the ADF test with a constant and trend, the Enders and Lee (2012a) test and the Omay (2015) test respectively. 10%, 5% and 1% critical values of the  $DF_{\tau}$  test are -3.131, -3.418 and 3.975; 10%, 5% and 1% critical values of the  $\tau_{DF-\tau}$  test for  $k = 2$  are -4.578, -3.985 and -3.676; 10%, 5% and 1% critical values of the  $\tau_{DF-\tau}^{fr}$  test at  $k^{fr}=1.70$  are -3.830, -4.140 and -4.700, respectively; \*\* (\*) indicate rejection of the null of unit root at 5% (10%) levels of significance.

can be seen from Figure A1, the fits are quite similar under the integer and the fractional cases, with both the estimates missing the sharp breaks during the ‘Great Depression’. Figure A2 provides an alternative picture of the fits in terms of the residuals, which tend to differ towards the latter half of the sample. As indicated by Omay (2015), it is reasonable to use only the 0.1 increments, because the other frequencies which are obtained for smaller increments are all clustered around this fractional frequency  $k^{fr} = 1.7$ , with the power loss not exceeding 1%. Hence, using this 0.1 increment for fractional frequency has the possibility of limiting type-two errors, over-filtration, and inappropriate nonlinear trend problems. In any event, more importantly, while the IFFFF and standard DF-type tests fail to reject the null hypothesis of unit root test, the more powerful FFFFF DF-type test indicates that real GNP of the US is, in fact, (non-linear) trend-stationary, and not difference-stationary as posited by Nelson and Plosser (1982).<sup>4</sup>

<sup>2</sup>In an earlier version of the Enders and Lee (2012a) paper, available at: <http://www3.nd.edu/~meg/MEG2004/Lee-Junsoo.pdf>, the authors also failed to reject the null of unit root in real GDP of the United States over the period of 1947:1–2003:2.

<sup>3</sup>Using standard unit root test with one break (Zivot and Andrews 1992) and two breaks (Lumsdaine and Papell 1997; Lee and Strazichic 2003) in both the mean and the trend, we could not reject the null of unit root even at 10% level of significance. Complete details of these results are available upon request from the authors.

<sup>4</sup>In a related area of research, many studies (Hamilton 1989; Perron 1989; Balke and Fomby 1991; Beaudry and Koop 1993; Murray and Nelson 2000; Kim, Morley, and Piger 2005; Camacho 2011; Hosseinkouchack and Wolters 2013) have accounted for the possibility that the persistence of the US output might differ in recessions and expansions. Not surprisingly, the evidence is, at best, mixed, with results depending on the tests and sample period used. Given this, following Hosseinkouchack and Wolters (2013), we tested the unit root hypothesis not only at the conditional mean of the real GNP, but also in the tails of the distribution using a quantile autoregression-based unit root test. Our results, based on the quantile Kolmogorov–Smirnov (QKS) test of Koenker and Xiao (2004), which checks for the unit root property across a range of quantiles, corroborated the findings of the Fractional Frequency Flexible Fourier Form Dickey and Fuller-type test. In other words, the QKS test confirmed that logarithm of the real GNP data (LGNP) is trend-stationary, with the results being driven by the quantiles ranging from 0.45 to 0.95 at 5% level of significance. Specifically, while recessions are found to be permanent, the results suggested that expansions are temporary. Next, when we filtered the LGNP data with the linear trend and the Fourier intercept, and then applied the quantile unit root test, the QKS test again confirmed that LGNP is trend-stationary. While stationarity is found to hold for quantiles 0.45–0.95 (at 5% level of significance) under  $k^{fr} = 2.0$ , the same holds for a bigger part of the distribution (0.35–0.95 at 5% level of significance) with  $k^{fr} = 1.667$ . In addition, we also used the  $DQ$  and  $SQ_{\tau}$  tests of Qu (2008) and Oka and Qu (2011) to check for breaks in the entire conditional distribution and at specific quantiles of the conditional distribution, respectively. For the entire conditional distribution, we obtained two breaks at 1903:2 and 1952:3, with the breaks primarily originating from the median of the distribution – the latter result being an indication of two regimes in the LGNP process. When we filtered the LGNP data with a linear trend and the two breaks (captured by dummy variables), we found that LGNP is trend-stationary over the quantile range of 0.35–0.95 at 5% level of significance. Complete details of these results are available upon request from the authors. Note that, our results are different from those obtained by Hosseinkouchack and Wolters (2013), whereby these authors indicated that null of unit root cannot be rejected over the entire conditional distribution of the US output (measured by real GDP over 1947:1–2012:1). However, these authors did not account for breaks in the conditional distribution of the real GDP.

## IV. Concluding remarks

This article aims to provide a definitive answer to the unresolved question of whether US output is best characterized by a trend- or difference-stationary process. To achieve our purpose, we apply the recently developed powerful unit root test with multiple smooth structural breaks of Omay (2015), based on a FFFFF on a unique data of US real GNP covering the quarterly period of 1875:1–2015:2. We find that while the standard Dickey and Fuller (1979) and the IFFFF of the DF test developed by Enders and Lee (2012a, 2012b) fail to reject the null of unit root, the FFFFF DF-type test provides strong evidence of the natural logarithm of real GNP as being trend-stationary. In other words, our article provides evidence that the US real GNP returns to a deterministic log-nonlinear trend (as characterized by smooth structural shifts) in the long run.

## Disclosure statement

No potential conflict of interest was reported by the authors.

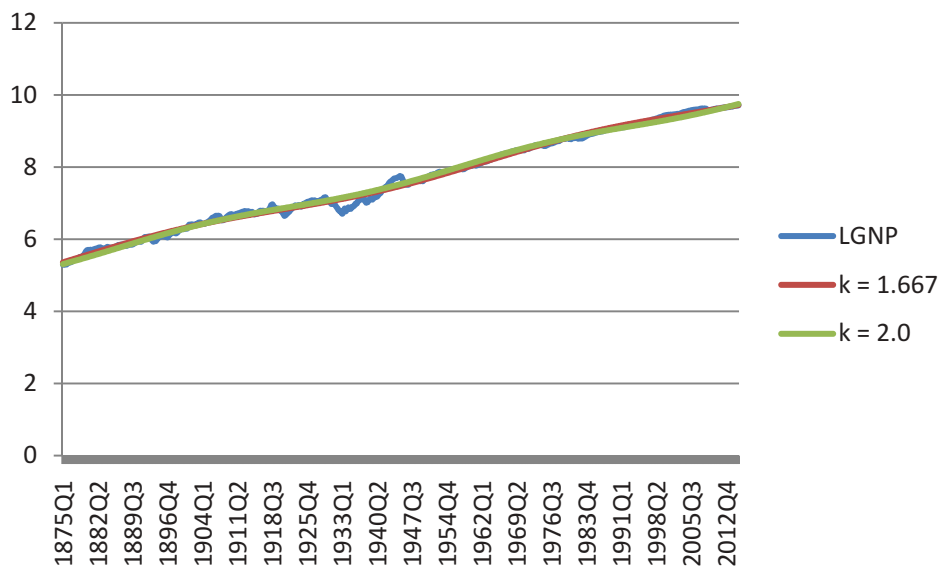
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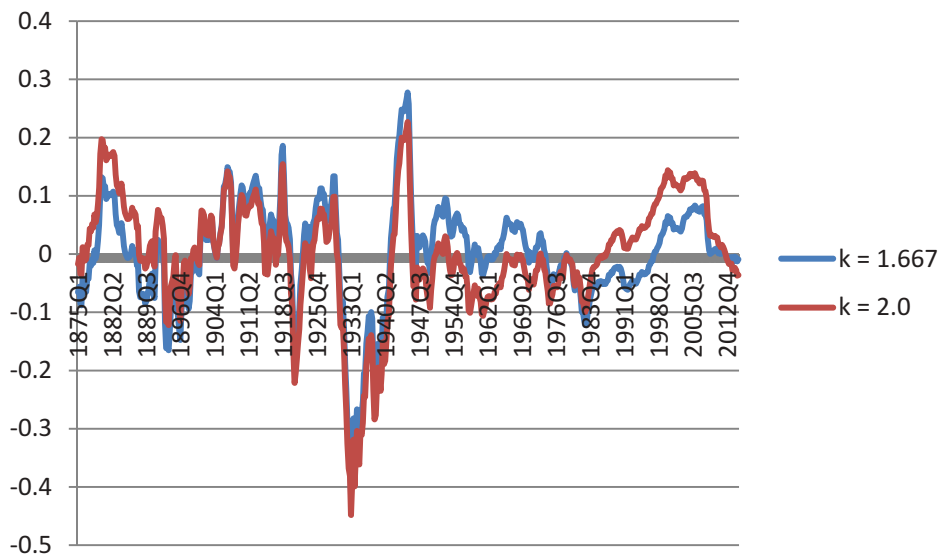
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## Appendix



**Figure A1.** Plot of the natural logarithms of US real GNP (LGNP) and fits under integer ( $k = 2.0$ ) and fractional ( $k = 1.667$ ) Fourier functions.



**Figure A2.** Plot of residuals after detrending with a linear trend and integer ( $k = 2.0$ ) and fractional ( $k = 1.667$ ) Fourier intercepts.