

This revision: May 23, 2003

Understanding Changes in International Business Cycle Dynamics

May 2003

James H. Stock

Department of Economics, Harvard University
and the National Bureau of Economic Research

and

Mark W. Watson*

Woodrow Wilson School and Department of Economics, Princeton University
and the National Bureau of Economic Research

*The authors thank Dick van Dijk, Brian Doyle, Jon Faust, Andrew Harvey, Siem Jan Koopman, and Graham Elliott for helpful comments and discussions. This research was funded in part by NSF grant SBR-0214131.

1. Introduction

Most of the G7 economies have experienced a reduction in the volatility of output growth, and a concomitant moderation of business cycle fluctuations, in the past two decades. Table 1 presents the decadal averages and standard deviations of four-quarter growth rates of per capita GDP for each of the G7 countries for the past four decades. Germany, Italy, Japan, the UK, and the US all experienced large reductions in volatility, while France and Canada experienced smaller moderations. Over this period, international trade flows have increased substantially, financial markets in developed economies have become increasingly integrated, and continental European countries moved to and eventually adopted a single currency. These developments raise the possibility of changes not only in the severity of international business cycles, but also in their synchronization.

There already is a large body of research on these changes, and there is a consensus concerning many of the basic facts. As initially pointed out by Kim and Nelson (1999) and McConnell and Perez-Quiros (2000), there has been a substantial moderation in output fluctuations in the US, with these and most other authors suggesting that this moderation is well modeled as a single break in the mid 1980s. Several explanations for this moderation have been proposed. Some of the proposed explanations, such as changes in monetary policy and adoption of new inventory management methods, are domestic in origin, while others, such as smaller international shocks or stabilizing effects of trade, have international roots; for further discussion and references see Blanchard and Simon (2001) and Stock and Watson (2002a).

Although the moderation of volatility is also evident in international data, when modeled as a single break the reductions generally are neither concurrent nor of similar magnitudes (e.g. Dalsgaard, Elmeskov and Park (2002), Del Negro and Otrok (2003), Doyle and Faust (2002), van Dijk, Osborn, and Sensier (2002), Fritsche and Kouzine (2003), Mills and Wang (2000), Simon (2001)). Moreover, existing research suggests little tendency towards increasing international synchronization of cyclical fluctuations (Doyle and Faust (2002a, 2002b), Heathcoate and Perri (2002), Kose, Prasad, and Terrones (2003)). Instead, there appears to have been an emergence of at least one

cyclically coherent group, the major countries in the Euro-zone (Artis, Kontelemis, and Osborn (1997), Artis and Zhang (1997, 1999), Carvalho and Harvey (2002), Helbling and Bayoumi (2003), Dalsgaard, Elmeskov and Park (2002), Del Negro and Otrok (2003), Luginbuhl and Koopman (2003)), and possibly a second, English-speaking group, consisting of Canada, the UK, and the US (Helbling and Bayoumi (2003)).

In light of this large literature, this paper has two specific objectives. The first is to provide a concise summary of the empirical facts about the moderation in output volatility, changes in persistence, and changes in cyclical comovements for the G7 countries. One conclusion is that the single-break model of variance reduction, which fits the US well, does not adequately describe the international patterns of moderation. In addition, we provide further evidence of the emergence of two cyclical groups, the Euro-zone and English-speaking groups.

Our second objective is to provide quantitative estimates of the sources of these changes: are they domestic or international in origin, and do they reflect changes in the magnitudes of structural shocks or, rather, changes in the response of the economies to those shocks? To obtain these estimates, we use a so-called factor-structural vector autoregression (FSVAR), specified in terms of the growth rates of quarterly GDP in the G7 countries. This FSVAR is a conventional structural VAR, where the identifying restrictions come from imposing an unobserved-component factor structure on the VAR innovations. The idiosyncratic shocks are allowed to affect future output in other countries, so this FSVAR makes it possible to quantify both the direct effect of common international shocks and the indirect effect of spillovers from the domestic shocks in one country to its trading partners. The FSVAR is overidentified, and tests of the overidentifying restrictions support the conclusion that the G7 output data are well described as being driven by two common international shocks, plus seven country-specific shocks. This FSVAR makes it possible to address various counterfactual questions, and (for example) makes it possible to estimate the extent to which the moderation in volatility is a result of smaller common international shocks, is domestic in origin, or is the result of a moderation in the US which spills over into the other G7 countries.

The data and the two methods we use to remove trends and to isolate business cycle components are briefly described in Section 2. Section 3 summarizes the empirical facts about changes in volatility and persistence for the individual G7 output data, and Section 4 summarizes the changes in international correlations. The FSVAR is described, and its overidentifying restrictions are tested, in Section 5. Empirical results and counterfactual calculations based on the FSVAR are presented in Section 6. Section 7 concludes.

2. Data and Filters

The data are quarterly per capita real GDP for the G7 countries (Canada, France, Germany, Italy, Japan, UK, and US), covering 1960:1 – 2002:4. Data sources are summarized in Table A-1.

The long-term growth rate of GDP is not constant for some of these countries, especially Germany, Japan, and Italy. The focus of this paper is fluctuations at yearly through business cycle horizons, not the determinants of postwar growth in Germany, Japan, and Italy. Because a low-frequency drift can introduce bias into certain statistics, such as cross-country correlations computed over large subsamples, we adopt two alternative methods for eliminating this low-frequency component.

The first method is to detrend the growth rate of GDP using an unobserved components model with a time-varying drift. Let $y_t = 400\Delta\ln(GDP_t)$ be the quarterly growth of GDP at an annual rate. We adopt an unobserved components specification which represents y_t as the sum of two terms, a slowly evolving mean growth rates and a stationary component:

$$y_t = \mu_t + u_t, \text{ where } \mu_t = \mu_{t-1} + \eta_t \tag{1}$$

and $a(L)u_t = \varepsilon_t$, where L is the lag operator and ε_t and η_t are serially and mutually uncorrelated mean zero disturbances. The Kalman smoother can be used to estimate the local mean, μ_t , and the residual. The detrended GDP growth rate is the residual, that is, the Kalman smoother estimate of u_t .

Implementing this detrending procedure requires a value of ratio $\sigma_\eta^2/s_{uu}(0)$, where $s_{uu}(0)$ is the spectral density of u_t at frequency zero. When $\sigma_\eta^2/s_{uu}(0)$, is small, as it is likely to be in this application, the maximum likelihood estimator of $\sigma_\eta^2/s_{uu}(0)$ has the “pileup” problem of having asymptotic point mass at zero even if its true value is nonzero but small, so we estimate $\sigma_\eta^2/s_{uu}(0)$ on a country-by-country basis using the median-unbiased estimator of Stock and Watson (1998), and use this country-specific estimate to detrend GDP growth.¹

The four-quarter growth rates of GDP in the seven countries are plotted in Figure 1. Figure 2 plots the detrended four-quarter growth rates, that is, the rolling four-quarter average of the detrended quarterly growth rates. Comparison of Figures 1 and 2 reveals that the detrending procedure eliminates the local mean of each series, but otherwise leaves the series essentially unchanged. Figure 2 also shows the simple average of the four-quarter growth rates for the G7 countries. Evidently many of these countries have episodes of considerable comovement, or synchronization, with aggregate G7 fluctuations.

The second filtering method we use is the Baxter-King (1999) band-pass (BP) filter, with 12 leads and lags and a pass-band of 8 to 32 quarters, applied to $\log(GDP_t)$. Band-pass-filtered GDP for each country are plotted in Figure 3, along with the average of the BP-filtered G7 GDP. Evidently BP-filtered GDP, like four-quarter growth, has periods of considerable international synchronization in business cycles. Notably, at the level of detail of Figures 2 and 3, the period of greatest synchronization appears to be the 1970s, and there is no readily apparent trend towards increased synchronization.

3. Changes in Volatility and Persistence

This section presents statistics summarizing the changes in the volatility of GDP and the persistence of innovations to GDP in the G7 countries.

¹ The median-unbiased estimates of $[T^2 \sigma_\eta^2/s_{uu}(0)]^{1/2}$ are: Canada, 3.8; France, 12.8; Germany, 5.4; Italy, 6.4; Japan, 13.2; UK, 0.0; and US, 3.3.

3.1 Volatility

As discussed in the introduction, there has been a substantial moderation in volatility in the post-1983 period, relative to the pre-1983 period. To get more detail on this moderation, we estimate the time path of the instantaneous variance of GDP using a nonGaussian smoother based on a stochastic volatility model with heavy variances and time-varying autoregressive coefficients. Let y_t be the quarterly GDP growth at an annual rate. The stochastic volatility model is,

$$y_t = \alpha_{0t} + \sum_{j=1}^p \alpha_{jt} y_{t-j} + \sigma_t \varepsilon_t, \text{ where } \alpha_{jt} = \alpha_{jt-1} + c \eta_{jt} \text{ and } \ln \sigma_t^2 = \ln \sigma_{t-1}^2 + \zeta_t, \quad (2)$$

where $\varepsilon_t, \eta_{1t}, \dots, \eta_{pt}$ are i.i.d. $N(0,1)$ and where ζ_t is distributed independently of the other shocks. To allow for large jumps in the instantaneous innovation variance, ζ_t is drawn from a mixture-of-normals distribution. The time-varying parameters were estimated by Markov Chain Monte Carlo methods. Given $\alpha_{0t}, \dots, \alpha_{pt}$, and σ_t^2 , it is possible to compute the instantaneous standard deviations of GDP growth, of four-quarter GDP growth, and of BP-filtered GDP for an idealized BP filter. For additional details, see Stock and Watson (2002a, Appendix A).

The resulting estimated instantaneous standard deviation of 4-quarter GDP growth is plotted in Figure 4. Different countries exhibit quite different paths of instantaneous standard deviations. In the US, there was a sharp moderation in the mid-1980s, while in the UK volatility declined in the late 1970s. Germany experienced a large but gradual decline in volatility, while volatility moderated in Japan but seems to have increased recently. Figure 5 plots the implied instantaneous standard deviation of BP-filtered GDP. Comparison of Figures 4 and 5 shows nearly identical patterns for four-quarter growth rates and BP-filtered GDP.

Formal tests for breaks in the conditional mean (that is, the autoregressive lag coefficients) and the conditional variance (that is, the autoregressive innovation variance)

of GDP growth are reported in Table 2.² The hypothesis of constant parameters is tested using the Wald version of the Quandt likelihood ratio (QLR) statistic, evaluated over the central 70% of the sample; the test of a constant conditional variance allows for the possibility of a break in the conditional mean at an unknown date that differs from the break date for the conditional variance. The break date and its 67% confidence interval are reported when the QLR statistic rejects at the 5% significance level. The final panel of Table 2 tests an alternative specification in which the innovation variance is modeled as a linear function of time with a discrete jump at an unknown break date, thereby nesting the single-break and linear time trend specifications.

The results in Table 2 indicate widespread instability in both the conditional mean and the conditional variance of these autoregressive models for GDP: in four of the seven countries, the hypothesis of a constant conditional mean is rejected at the 5% level, and in all countries but Japan the hypothesis of a constant conditional variance is rejected. For the US, the results in the final panel suggest that the break model is preferred to a linear time trend: in the nested specification, the break is significant but the time trend is not. This finding does not generalize to the other countries, however. For example, for Italy neither the trend term nor the break term are individually significant in the nested specification. This finding does not imply that the variance for Italy was constant, for the test in panel B rejects the no-break specification at the 1% level and the estimated instantaneous variances in Figure 3 indicates a substantial reduction in volatility over this period; rather, the nonrejections for Germany, Italy, and Japan – and the significance of both terms for the UK – suggests that neither the single-break nor the linear-decline model provides a good summary of the changing volatility for these countries. Rather, for most G7 countries, the pattern of change in the volatility of GDP is more complex.

3.2 Persistence and Size of Univariate Shocks

² Raw (i.e. not detrended) GDP growth rates are used in the tests in Table 2 to ensure that currently available theory can be used for the null distribution of the break tests (Andrews (1991)) and the distribution of the break date estimators (Bai (1997)).

Another way to look at the changing autocovariances of GDP growth in these countries is to examine changes in the variance of the AR innovation and in the sum of the AR coefficients, which measures the persistence of a shock to GDP growth. Changes in the variance of GDP growth imply that its spectrum has changed; an increase in the sum of the AR coefficients implies an increase in the relative mass at frequency zero, while a change in the innovation variance implies a shift in the level (but not necessarily the shape) of the spectrum.

Table 3 presents the sum of the coefficients and the one-step ahead forecast standard error for univariate AR models of detrended GDP growth. The sum measures the persistence of an innovation to GDP, and by this measure GDP innovations have become substantially more persistent for Canada, France, and the UK. Persistence has not changed much for the US and Italy, while it has declined for Germany and Japan. For all countries, the magnitude of the GDP innovations, as measured by the standard error of the regression, has decreased substantially: one-quarter ahead forecasts based on univariate autoregressions have become more accurate for the G7 countries.

Figure 6 presents estimated time paths for the sum of coefficients and AR error variance. These estimates are computed using a rolling fourth-order autoregression, where the autoregression is estimated by weighted least squares with two-sided exponential weighting, with quarterly discount factor of .97; the date plotted in the figure corresponds to the central observation with the greatest weight.³ These figures are generally consistent with the two-sample evidence in Table 3. In all countries, the innovation variance has fallen substantially, although it increased again during the 1990s in Japan. In Canada, France, and the UK, persistence has increased substantially, while persistence has been roughly constant for the US. The timing of these changes differs across countries, a result consistent with variations in the patterns of declining variances in Figures 4 and 5.

³ That is, the estimate plotted at date s is based on weighted least squares estimation of the AR using all observations, with the observation at date t receiving weight $.97^{|s-t|}$. Similar results are obtained using the nonGaussian smoother estimates based on (2). The two-sided exponential weighting scheme is used here for comparability with the two-sided VAR estimates reported in Sections 4 and 6.

4. Changes in Synchronization

This section reports various measures of time-varying international comovements of GDP. To facilitate comparability with the analysis of Sections 5 and 6 using the FSVAR, these measures are estimated using a reduced form seven-country VAR. The section begins by describing the reduced form VAR, then turns to the measures of time-varying correlations.

4.1 Reduced Form VAR

A conventional VAR(p) with all seven countries would have $7p$ coefficients in each equation, where p is the number of lags. With the short quarterly data set at hand, this many coefficients would induce considerable sampling uncertainty even with small values of p . One solution to this dimensionality problem is to consider VARs specified in terms of subsets of countries, as is done by Doyle and Faust (2002b), but this limits the international spillovers and common shocks that can be studied in a single model. Another solution is to specify a model for all seven countries but to impose additional restrictions on the VAR coefficients, as is done in many papers in this literature, for example Helg et. al. (1995). In this paper, we consider two such sets of restrictions.

For the main results, the restriction we use is for lagged foreign GDP growth to enter with a different number of lags than domestic GDP growth. Specifically, let Y_t be the vector of detrended quarterly GDP growth rates. The reduced form VAR is

$$Y_t = A(L)Y_{t-1} + v_t, \text{ where } E v_t v_t' = \Sigma \quad (3)$$

where the diagonal elements of the matrix lag polynomial $A(L)$ have degree p_1 and the off-diagonal elements have degree p_2 . Denoting the resulting VAR by VAR(p_1, p_2), we used the AIC and BIC to compare various p_1 and p_2 in the two subsamples and the results pointed to a VAR(4,1) specification.

The second restriction we considered further restricts the coefficients on the lags of foreign GDP to be proportional to their trade shares, an approach taken by Elliott and

Fatás (1997) and Norrbin and Schlagenhauf (1996). We examine results from that VAR as part of the sensitivity analysis in Section 6.

The second moments of interest in this paper can all be computed directly from estimates of the VAR parameters in (3). The spectral density matrix of quarterly growth Y_t is $S_{YY}(\omega) = A(e^{i\omega})\Sigma A(e^{-i\omega})'/2\pi$. The implied spectral density matrix of four-quarter GDP growth is $\|1 + e^{i\omega} + e^{2i\omega} + e^{3i\omega}\|^2 S_{YY}(\omega) = \{s_{ij}^{(4)}(\omega)\}$, so that $s_{ij}^{(4)}(\omega)$ is the cross-spectrum (spectrum when $i = j$) between four-quarter GDP growth in country i and country j at frequency ω . The implied spectral density matrix of BP-filtered GDP growth is $\|b(e^{i\omega})/(1 - e^{i\omega})\|^2 S_{YY}(\omega)$, where $\|z\|^2$ is the squared norm of the possibly complex variable z and b is the idealized BP filter so that $\|b(e^{i\omega})\|^2 = 1$ for $\omega_0 \leq \omega \leq \omega_1$, where the frequencies ω_0 and ω_1 respectively correspond to periodicities of 32 and 8 quarters. Thus, for example, the contemporaneous correlation $\rho_{ij}^{(4)}$ between four-quarter growth rates in countries i and j is

$$\rho_{ij}^{(4)} = \frac{\int_{-\pi}^{\pi} s_{ij}^{(4)}(\omega) d\omega}{\left(\int_{-\pi}^{\pi} s_{ii}^{(4)}(\omega) d\omega\right)^{1/2} \left(\int_{-\pi}^{\pi} s_{jj}^{(4)}(\omega) d\omega\right)^{1/2}}. \quad (4)$$

Various second moments were computed both for the two subsamples and for rolling estimates of the VAR parameters. The rolling VAR parameters were estimated by weighted least squares using the two-sided exponential weighting scheme described in Section 3 for the rolling ARs.

4.2 International Synchronization

Table 4 presents various measures of international output comovements. Panels A and B tabulates the correlation of four-quarter GDP growth rates across countries, first using the raw data then based on the estimated model. The average absolute difference of these estimated correlations is .04 in the first subsample and .11 in the second subsample, indicating that that the reduced form VAR(4,1) captures most of the business cycle

comovements of these series; the biggest exception is that the VAR(4,1) estimated correlation considerably exceeds the sample correlation between US and French four-quarter GDP growth in the second period. Panel C of Table 4 presents the correlations among BP filtered GDP estimated using the reduced form VAR(4,1); the entries in Panel C and correlations estimated directly from BP-filtered data (not tabulated) differ by an absolute average of .06 in the first period and .11 in the second period.

Rolling correlations between own-country BP-filtered GDP and US and German BP-filtered GDP, based on the reduced-form VAR(4,1), are plotted in Figure 7; like the other plots of rolling estimates, the plotted date corresponds to the center of the rolling window, the date with the greatest weight in the rolling exponential weighting scheme.

Three aspects of Table 4 and Figure 7 bear emphasis. First, as emphasized by Doyle and Faust (2002a, 2002b), Heathcoate and Perri (2002), and Kose, Prasad, and Terrones (2003), there is no overall tendency towards closer international synchronization over this period: depending on the correlation measure used, the average cross-country correlation either is unchanged between the two subsamples or drops slightly.

Second, there nevertheless appears to have been substantial shifts in comovements among the G7 economies, but these shifts are more towards the emergence and evolution of Euro-zone and English-speaking regional groups. Based on the correlations in Table 4(a), during the first subsample the average correlation within the two groups was .49 (continental Europe) and .51 (English-speaking), and the average cross-group correlation was .42. In contrast, in the second period the average correlations within the two groups rise to .63 (continental Europe) and .71 (English-speaking), while the average cross-group correlation drops to .29. A significant part of this change is the decline in the correlation between UK GDP growth and that of France and Germany, and an increase in its correlation with the North American economies. The emergence of the two regional groups, English-speaking and Euro-zone, also is evident in Figure 7 through the increasing correlation between France and Italy with Germany and the increasing correlation between the UK and the US (and its decreasing correlation with Germany).

Third, the synchronization of Japanese cycles has been low throughout this forty-year period and recently decreased further. Based on the correlations of four-quarter

GDP growth in Table 4(a), from 1960 – 1983 the average correlation between Japan and the other G7 countries fell from 0.35 during 1960 – 1983 to 0.15 during 1984 – 2002.

5. The Factor-Structural VAR model

There are several frameworks available for developing a time series model with enough structure to permit answering the questions of interest here, such as the fraction of a countries cyclical variance that is due to international shocks and how that has changed over time. Before discussing the specific framework used in this paper, a structural VAR, it is useful to discuss the various modeling options available and to assess their strengths and weaknesses.

The basic issue to be resolved is the best way to identify a world (or G7) shock. One approach is simply to define a world shock to be the innovation to world (or G7) GDP growth. While this approach has the advantage of being easy to implement, because US output receives great weight in G7 GDP it confounds world shocks with US shocks and idiosyncratic shocks to other large economies. Suppose there were in fact no common shocks and no trade; this identification scheme would nevertheless attribute a large fraction of US fluctuations to a common shock as an arithmetic implication of its construction.

A second approach is to implement a parametric dynamic factor model in which the number of shocks exceeds the number of series, and the comovements across series at all leads and lags are attributed to the common shock. This leads to an unobserved components model that can be estimated using Kalman filtering and related methods. This approach has been widely used in the international fluctuations literature; recent contributions include Kose, Otrok, and Whiteman (2001), Kose, Prasad, and Terrones (2003), Harvey and Carvalho (2002), Luginbuhl and Koopman (2003), and Monfort, Renne, Ruffer, and Vitale (2002). This framework has several advantages. In the hypothetical case of no trade and no common shocks, there would be no comovements and the common shock would correctly be estimated to have zero variance. This framework also captures the differences in dynamic responses of different economies to a world shock. On the other hand, because cross-dynamics are all attributed to the world

shock, this approach is not well suited to separately identifying the effects of a common world shock and spillovers arising through trade: if there were in fact no world shocks but idiosyncratic shocks were transmitted through trade, the parametric dynamic factor model would incorrectly estimate a nonzero world shock.⁴

A third approach is to use nonparametric methods to estimate a dynamic factor model. If a large number of series have a dynamic factor structure, then the common component or the common dynamic factor can be estimated using principal components (Stock and Watson (2002b)) or dynamic principal components (Forni, Hallin, Lippi, and Reichlin (2000)). This strategy is used by Helg et. al. (1995) to extract European industry and country shocks as principal components of reduced-form VAR errors, and by Helbling and Bayoumi (2003) to estimate the importance of common factors in G7 fluctuations. Prasad and Lumsdaine (2003) also adopting this strategy, using a weighting scheme rather than principal components to extract the innovation in a single common G7 factor. In principle the principal components/nonparametric approach has the advantages of the second approach without the disadvantage of assuming that all comovements stem from the common disturbance rather than through trade spillovers; in practice, however, if this approach is based on only the G7 then individual countries are necessarily heavily weighted leading to the same problems as the first approach, in particular finding a common factor even if there is none.

A fourth approach, the one used here, is to adopt a VAR framework for the lagged effects but to identify world shocks as those that affect all countries within the same period. Thus country-specific shocks can lead to spillovers, but those spillovers are assumed to happen with at least a one-quarter lag. In practice this results in an overidentified structural VAR, in which the shocks are identified by imposing a factor structure on the reduced-form errors. Examples of papers using this approach (in a regional or international context) include Altonji and Ham (1990), Norrbin and Schlagenhauf (1996), and Clark and Shin (2000). Two advantages of this approach are

⁴ Monfort, Renne, Ruffer, and Vitale (2002) partially address this drawback by considering, as an alternative to their main analysis, a specification with regional shocks that interact dynamically and thus allow cross-region spillovers. Going further down this route and fully relaxing the lag dynamics would lead to the FSVAR model discussed below.

that no world shock is estimated if there is none, or if all cross-dynamics arise from lagged trade spillovers. A potential disadvantage of this approach is that if an international shock affects several countries only with a lag, that effect may incorrectly be interpreted as a spillover.

We therefore consider the FSVAR model consisting of the VAR model (3) in which the errors have the factor structure,

$$v_t = \Gamma f_t + \xi_t, \text{ where } E(ff_t') = \text{diag}(\sigma_{f_1}, \dots, \sigma_{f_k}) \text{ and } E(\xi_t \xi_t') = \text{diag}(\sigma_{\xi_1}, \dots, \sigma_{\xi_7}), \quad (5)$$

where f_t are the common international factors, Γ is the $7 \times k$ matrix of factor loadings, and ξ_t are the country-specific, or idiosyncratic, shocks. In (5), common international shocks or factors are identified as those shocks that affect international output contemporaneously.

The FSVAR specification (5) is overidentified, so that empirical evidence can be brought to bear on the number of factors k . Tests of the overidentifying restrictions are summarized in Table 5. In both subsamples and in the pooled full sample, the hypothesis of $k=1$ is rejected against the unrestricted alternative (that is, against Σ_v having full rank) at the 1% significance level, but the null hypothesis of $k=2$ is not rejected at the 20% significance level. The testing results in Table 5 clearly support a specification with two common international shocks.

For the decompositions and counterfactual calculations in Sections 6.2 and 6.3, we need to adopt a further normalization to identify separately the factor loadings Γ and the variances of the common shocks. The specific normalization we adopt is that the first factor has contemporaneous unit effect on US GDP, that is, $\Gamma_{US,1} = 1$; and that the second factor has immediate unit effect on France but does not immediately affect the US, so $\Gamma_{France,1} = 1$ and $\Gamma_{US,2} = 0$.

6. Empirical Results

This section presents empirical results based on the FSVAR, including an analysis of the sensitivity of the results to some modeling decisions.

6.1 Changing Importance of Common and Country-Specific Shocks

The factor structure permits a decomposition of the h -step ahead forecast error for GDP growth in a given country into three sources: unforeseen common shocks, unforeseen domestic shocks, and spillover effects of unforeseen domestic shocks to other G7 countries. Because the country shocks and the common shocks are all uncorrelated by assumption, this decomposition in turn permits a threefold decomposition of the variances of the h -step ahead forecast error. This approach also can be used to decompose the variance of BP-filtered GDP.

Table 6 summarizes these variance decompositions for GDP growth and for BP-filtered GDP. At the one-quarter horizon, international spillovers account for none of the GDP growth forecast error variance: this is the assumption used to identify the international shock. At longer horizons, spillovers account for between 5% and 20% of the variance of GDP growth, depending on the country and the subsample. Most of the variance of GDP growth is attributed to the common shocks or to idiosyncratic domestic shocks; the breakdown varies considerably across countries, however. In the first period, the effects of international shocks at the four-quarter horizon are estimated to be the greatest for Canada and France, and the least for Japan and Italy. In the second period, almost all the forecast error variance in Japan is attributed to domestic shocks, a result consistent with the declining correlation between GDP in Japan and in other countries in the second period reported in Section 4.

One measure of the degree of exposure of domestic economies to international sources of fluctuations, either common shocks or spillovers, is the share of the forecast error variance attributed to domestic shocks; a small domestic corresponds to a relatively larger role for international rather than domestic disturbances. By this measure, the international exposure of Canada, France, and Italy has increased substantially, the exposure of the US has increased moderately, that of Germany and the UK has decreased

moderately, but the effect on Japan of international linkages has dropped to less than 10%. The variance decompositions for BP-filtered GDP yield similar conclusions to the variance decompositions of GDP growth at the 4- and 8-quarter ahead horizon.

Figure 8 presents time-varying estimates of the variance decomposition of BP-filtered GDP, based on rolling estimates FSVAR (as before, using exponential weighting). The units in Figure 8 are those of the variance; the lower line is the contribution to the variance of the international shocks, the middle line is the sum of the contributions of the international shocks and spillovers, and the top line is total variance, so the gap between the top and middle lines is the contribution to the variance of domestic shocks. For Canada, Germany, the UK, and the US, the decline in the overall volatility tracks a decline in the variance arising from international shocks. For Italy, the large historical decline in the variance is associated with a declining importance of domestic shocks. For Japan, international shocks have become unimportant, and domestic shocks explain nearly all of its volatility in the 1990s and are the source of its recent increase in volatility.

6.2 Changes in Volatility: Impulse or Propagation?

In principle, the contribution of international shocks to output volatility could decrease because the variance of the international shocks has decreased, because a shock of a fixed magnitude has less of an effect on the economy, or both. Said differently, the variance of GDP growth in a given country can change because the magnitude of the shocks impinging on that economy have changed or because the effects of those shocks have changed.

In this section, we decompose the change in the variance from the first subsample to the second into changes in the magnitudes of the shocks (“impulses”) and changes in their effect on the economy (“propagation”). To make this precise, let V_p denote the variance of the 4-quarter-ahead forecast errors in a given country in period p , where $p = 1, 2$ corresponds to 1960 – 1983 and 1984 – 2002. The variance decomposition attributes a portion of V_p to each of the nine shocks in the model, so we can write, $V_p = V_{p,1} + \dots + V_{p,9}$, where $V_{p,j}$ is the variance in period p attributed to the j^{th} shock. Thus the change in the variance between the two periods is $V_2 - V_1 = (V_{2,1} - V_{1,1}) + \dots + (V_{2,9} - V_{1,9})$. In

identified structural VARs, the variance component $V_{p,j}$ always can be written as $a_{pj}\sigma_{pj}^2$, where a_{pj} is a term depending on the squared impulse response of GDP to the j^{th} in period p and σ_{pj}^2 is the variance of shock j in period p . Thus the change in the contribution of the j^{th} shock can be decomposed exactly as,

$$V_{2,j} - V_{1,j} = \left(\frac{a_{1j} + a_{2j}}{2} \right) (\sigma_{2j}^2 - \sigma_{1j}^2) + \left(\frac{\sigma_{1j}^2 + \sigma_{2j}^2}{2} \right) (a_{2j} - a_{1j}). \quad (6)$$

That is, the change in the variance can be decomposed into the contribution from the change in the shock variance plus the contribution from the change in the impulse response. The decomposition (6) is additive so these contributions can be aggregated into variance changes arising from the common shocks, spillovers, and own shocks, with each type of shock in turn decomposed into changes in variances arising from changing shock variances and from changing impulse responses; this yields a six-way decomposition of the change in the variance of GDP forecast errors from the first period to the second.

Table 7 presents this six-way decomposition of the change in variances of four quarter-ahead forecast errors in GDP. Evidently, the decline in the variance between the two periods is to a great extent attributed to a decline in the magnitudes of the shocks. Indeed, for Canada, France, the UK, and to a lesser extent the US, the decline in the shock variances more than accounts for the drop in the variance of GDP forecast errors, in the sense that changes in the propagation mechanism worked to increase rather than decrease the total variance across these two periods. For Germany and Italy, the net contribution of changes in propagation is small, so that most of the variance reductions in Germany and Italy are attributed to changes in the magnitudes of the shocks. The exception here, as we have seen in other aspects of this analysis, is Japan, in which the decline in the variance is largely attributed to changes in the propagation mechanism, not to changes in the size of shocks. Among the different types of shocks, reductions in the size of international shocks played a substantial role in the volatility moderation in all countries but Japan (where the international shock is estimated to have negligible effect).

In addition, in all countries a small but nonzero portion of the moderation in all is attributed to smaller foreign idiosyncratic shocks.

One lesson from Table 7 is that there have been important changes in the effect of an international shock of a fixed magnitude on some of these economies. This changing effect is examined further in Figure 9, which presents the impulse response functions for the different countries in the two subsamples with respect to the first common factor (Figure 9a) and the second common factor (Figure 9b). The most noticeable feature is that for many countries, in particular the Canada, France, the UK, and the US, there is a large estimated increase in the magnitude of the effect of the common shocks and in their persistence. Again, Japan is different than the rest of the G7, with the estimated responses to both shocks being nonzero in the first period but nearly zero in the second.⁵

6.3 Counterfactuals: Second Period Propagation, First Period Shocks

The foregoing results indicate that much of the moderation is attributable to declines in the variance of the common international shocks. This raises the counterfactual question: what would the volatility and cross-correlations have been in 1984 – 2002, had the G7 economies been confronted with common international shocks as large as they experienced in 1960 – 1983?

This counterfactual question can be addressed by suitably combining the impulse responses from the second period FSVAR with the shock variances from the first period FSVAR, then computing the implied moments. The resulting estimated variances are summarized in Table 8 for 4-quarter GDP growth and BP-filtered GDP growth. Comparing the first line of each panel (the estimated standard deviations based on second-period impulse response functions and second-period shock variances) with the second line (in which the first-period variance of the common shocks is used) reveals that all countries, except again Japan, would have had considerably greater volatility over the past two decades had the world experienced the first-period shocks. For example, the

⁵ The impulse response functions in Figure 9 apply to the arbitrary normalization of the two common shocks discussed at the end of Section 5. Although different normalizations would yield different impulse response functions, they would be linear combinations of the impulse response functions in Figure 9, and much of the discussion about persistence in this paragraph would apply to those linear combinations as well.

standard deviation of four-quarter GDP growth in the US would have been approximately 3.2 percentage points, compared with the actual value of 1.7 percentage points; the standard deviation of French four-quarter GDP growth, which in reality was essentially constant over the two periods, would have doubled had the second period experienced international shocks of the same magnitude as the first period.

The cross-country correlations implied by this counterfactual scenario are summarized in Table 9. Not surprisingly, had the common shocks been larger in the second period than they actually were, the cross-country correlations would have been higher, and indeed would have been much higher than they were in the first period. Under the counterfactual scenario the correlations typically increase by between .10 and .20 (Japan again being the exception). According to these estimates, had the common shocks in the second period been as large as they were during the first period, international business cycles would have been more highly synchronized than they actually were, and indeed would have been more highly synchronized than there were in the 1960 – 1983 period.

6.4 Sensitivity analysis

This section reports the results of two checks of the foregoing results to changes in the modeling assumptions or in the precise statistics reported.

Trade-weighted VAR lag restrictions. As a check, we considered a further restriction of the VAR in which the coefficients on foreign GDP are proportional to trade shares. Elliott and Fatás (1996) used a similar restriction to identify shocks in a structural VAR, and Norrbin and Schlagenhauf (1996) used it (as we do here) to simplify the lag dynamics. Accordingly, the restricted reduced form VAR is,

$$Y_t = b(L)Y_{t-1} + d(L)WY_{t-1} + v_t, \text{ where } Ev_t v_t' = \Sigma \quad (7)$$

where $b(L)$ and $d(L)$ are 7×1 vector lag polynomials and W is a fixed weighting matrix, where the diagonal elements of W are zero and the (i,j) element is the share of gross trade (imports plus exports) of trading partner j in all of country i 's trade with G7 countries.

In the restricted reduced form VAR (7), the number of coefficients per equation equals the number of own lags (the degree of $b(L)$) plus the number of lags on trade-weighted foreign GDP (the degree of $d(L)$). AIC and BIC comparisons point to 4 own lags and 1 lag of trade-weighted foreign GDP growth. The FSVAR corresponding to (7) imposes the factor structure (5) on the reduced form errors in (7).

As a gauge of the sensitivity of the results in the previous sections, we recomputed the counterfactual variances and correlations of Tables 8 and 9 for the trade-weighted FSVAR; the results are reported in Tables 10 and 11. Although the numerical values for the estimated changes in variances in Tables 8(a) and 10 differ, the qualitative conclusions are the same. In all countries, the variances of four-quarter GDP growth would have been considerably larger had second-period shocks been as large as first-period shocks. The main difference between the standard deviations in Table 8(a) and Table 10 is the estimated increase for France, which is less using the trade-weighted FSVAR than the base case FSVAR in Table 8(a). The main differences between the implied correlations in Tables 9 and 11 is that the model-based estimates in Table 11(a) (estimates of actuals, not counterfactuals) in some cases differ considerably from those in Table 9(a) and from the sample correlations in Table 4(a); in this sense at least, the trade-weighted FSVAR does not fit the data as well as the base case FSVAR. However, the main conclusion from Table 9 – that international synchronization would have been substantially greater had the common shocks in the second period been as large as they were in the first – is also obtained using the trade-weighted FSVAR, so the main conclusions from the counterfactual analysis are robust to using this additional restriction.

Measuring synchronization by average coherences. The analysis of international synchronization so far has relied on contemporaneous cross-correlations of four-quarter GDP growth and of BC-filtered GDP as the measures of comovements, but this can mask lagged associations. An alternative measure of comovements, which is invariant to these lagged effects, is the average coherence at business cycle frequencies. Specifically, let ω_0 and ω_1 be the lower and upper frequencies that define the business cycle portion of the spectrum, and let $s_{ij}^{(4)}(\omega)$ be the cross-spectrum between the four quarter growth rates in countries i and j as in (4). One measure of the average coherence between four-quarter growth rates in countries i and j at business cycle frequencies is,

$$R_{ij}^2(\omega_0, \omega_1) = \frac{\int_{\omega_0}^{\omega_1} \|s_{ij}^{(4)}(\omega)\|^2 d\omega}{\left(\int_{\omega_0}^{\omega_1} \|s_{ii}^{(4)}(\omega)\|^2 d\omega \right)^{1/2} \left(\int_{\omega_0}^{\omega_1} \|s_{jj}^{(4)}(\omega)\|^2 d\omega \right)^{1/2}}. \quad (8)$$

This measure reduces to the usual definition of the coherence when it is evaluated at a single frequency rather than over the range ω_0 to ω_1 .

The square root of average coherence (8), $R_{ij}(\omega_0, \omega_1)$, was computed for the counterfactual correlations examined in Table 9, and the results are summarized in Table 12 (because the coherence has the interpretation of an R^2 , using the square root of the average coherence makes this measure more directly comparable to the correlations of Table 9). Comparing panel (a) of Tables 9 and 11 shows that the coherences are slightly higher than the correlations of four-quarter growth rate, which is not surprising because the coherences are not sensitive to phase shifts and also focus on business cycle frequencies, whereas the four-quarter growth rates have some higher frequency noise. The qualitative conclusions from the counterfactual exercise, however, are the same as those drawn from Table 9: under the counterfactual scenario, average business cycle coherences increase, typically by .10 or .20; the exception remains Japan, where the average coherences increase by less. In general, findings based on the contemporaneous correlations and the average coherence will be different. As it happens, however, the cross-country lead-lag relations are not large, so the phase shift is small and these different measures give similar results.

7. Discussion and Conclusion

These empirical results suggest four broad conclusions. First, although there has not been a general increase in international synchronization among G7 business cycles, there have been important changes, in particular the emergence of two groups, one consisting of Euro-zone countries and the other of English-speaking countries, within which correlations have increased and across which correlations have decreased. Over

this period, cyclical movements in the UK became less correlated with Euro-zone countries and more correlated with North American countries.

Second, common international shocks have been smaller in the 1980s and 1990s than they were in the 1960s and 1970s. According to the FSVAR, this declining volatility of common G7 shocks is the source of much of the observed moderation in individual country business cycles. Moreover, this moderation of common G7 shocks is responsible, in a mechanical sense, for the failure of business cycles to become more synchronous as one might expect given the large increase in trade over this period: had world shocks been as large in the 1980s and 1990s as they were in the 1960s and 1970s, international cyclical correlations would have increased considerably.

Third, the Japanese experience is in many ways exceptional. For the other G7 countries, volatility generally decreased or at least stayed constant in the 1990s, but it increased in the 1990s in Japan. During the 1980s and 1990s, cyclical fluctuations in Japanese GDP became almost detached from the other G7 economies, with domestic shocks explaining almost all of the cyclical movements in Japanese GDP.

Fourth, a robust finding is that, however measured, persistence of disturbances – both reduced-form innovations and structural shocks – has increased in Canada, France, the UK. In those countries, a shock of a given magnitude would result in more cyclical volatility today than thirty years ago.

This analysis has focused on documenting the changes in the magnitudes of shocks and their effects. An important next step is sorting out the reasons for these changes and their implications for economic policy.

References

- Ahmed, S., A. Levin and B. Wilson (2001), "Recent U.S. Macroeconomic Stability: Good Luck, Good Policies, or Good Practices?" manuscript, FRB BOG.
- Andrews, Donald W.K. (1993), "Tests of Parameter Instability and Structural Change with Unknown Change Point," *Econometrica*, 61, 821-856.
- Artis, M.J., Z.G. Kontolemis, and D.R. Osborn (1997), "Business Cycles for G7 and European Countries," *Journal of Business* 70, 249 – 279.
- Artis, M.J. and W. Zhang (1997), "International Business Cycles and the ERM: IS there a European Business Cycle," *International Journal of Finance and Economics* 2, 1 – 16.
- Artis, M.J. and W. Zhang (1999), "Further Evidence on the International Business Cycle and the ERM: Is there a European Business Cycle?" *Oxford Economic Papers* 51, 120 – 132.
- Bai, J. (1997), "Estimation of a Change Point in Multiple Regression Models," *Review of Economics and Statistics* 79, 551-563.
- Blanchard, O. and J. Simon (2001), "The Long and Large Decline in U.S. Output Volatility," *Brookings Papers on Economic Activity* 2001:1, 135 – 164.
- Carvalho, V.M. and A.C. Harvey (2002), "Convergence and Cycles in the Euro-Zone," manuscript, University of Cambridge
- Clark, T. and K. Shin (2000), "The Sources of Fluctuations Within and Across Countries," in G. Hess and E. van Wincoop (eds.), *Intranational Macroeconomics*, Cambridge, UK: Cambridge University Press.
- Dalgaard, T., J. Elmeskov, and C-Y Park (2002), "Ongoing Changes in the Business Cycle – Evidence and Causes," OECD Economics Department Working Paper no. 315.
- Del Negro, M. and C. Otrok (2003), "Time-Varying European Business Cycles," manuscript, University of Virginia.
- van Dijk, D., D.R. Osborn, and M. Sensier (2002), "Changes in the Variability of the Business Cycle in the G7 Countries," manuscript, Erasmus University Rotterdam

- Doyle, B. and J. Faust (2002a), "An Investigation of Co-movements among the Growth Rates of the G-7 Countries," *Federal Reserve Bulletin*, October 2002, 427 – 437.
- Doyle, B. and J. Faust (2002b), "Breaks in the Variability and Co-Movement of G-7 Economic Growth," manuscript, Federal Reserve Board.
- Elliott, G. and A. Fatás (1996), "International Business Cycles and the Dynamics of the Current Account," *European Economic Review* 40, 361 – 387.
- Forni, M., M. Hallin, M. Lippi, and L. Reichlin (2000), "The Generalized Dynamic Factor Model: Identification and Estimation," *The Review of Economics and Statistics* 82, 540-552.
- Fritsche, U. and V. Kouzine (2003), "Declining Output Volatility and Monetary Policy in Germany," manuscript, German Institute for Economic Research (DIW Berlin).
- Heathcoate, J. and F. Perri (2002), "Financial Globalization and Real Regionalization," NBER Working Paper 9292.
- Helbling, T. and T. Bayoumi (2003), "Are They All in the Same Boat? The 2000-2001 Growth Slowdown and the G-& Business Cycle Linkages," manuscript, International Monetary Fund.
- Helg, R., P. Manasse, T. Monacelli, and R. Roveli (1995), "How Much (A)symmetry in Europe? Evidence from Industrial Sectors," *European Economic Review* 39, 1017 – 1041.
- Kim, C.-J. and C.R. Nelson (1999), "Has the U.S. Economy Become More Stable? A Bayesian Approach Based on a Markov-Switching Model of the Business Cycle," *The Review of Economics and Statistics* 81, 608 – 616.
- Kose, M.A., C. Otrok, and C.H. Whiteman (2001), "International Business Cycles: World, Region, and Country-Specific Factors," forthcoming, *American Economic Review*.
- Kose, M.A., E.S. Prasad, and M.E. Terrones (2003), "How Does Globalization Affect the Synchronization of Business Cycles?," IZA Discussion Paper no. 702.
- Luginbuhl, R. and S.J. Koopman (2003), "Convergence in European GDP Series: A Multivariate Common Converging Trend-Cycle Decomposition," Tinbergen Institute Discussion Paper 2003-031/4.

- McConnell, M.M and G. Perez-Quiros, (2000), "Output Fluctuations in the United States: What has Changed Since the Early 1980's," *American Economic Review*, Vol. 90, No. 5, 1464-1476.
- Mills, T.C. and P. Wang (2000), "Searching for the Sources of Stabilisation in Output Growth Rates: Evidence from the G-7 Economies," manuscript.
- Monfort, A., J.P. Renne, R. Ruffer, and G. Vitale (2002), "Is Economic Activity in the G7 Synchronized? Common Shocks vs. Spillover Effects," manuscript.
- Norrbin, S.C. and D.E. Schlagenhauf (1996), "The Role of International Factors in the Business Cycle: A Multi-Country Study," *Journal of International Economics* 40, 84 – 104.
- Prasad, E.S. and R.L. Lumsdaine (2003), "Identifying the Common Component of International Economic Fluctuations: A New Approach," *Economic Journal* 113, 101 – 127.
- Simon, J. (2001), "The Decline in Australian Output Volatility," manuscript, Reserve Bank of Australia
- Stock, J.H. and M.W. Watson (1998), "Asymptotically Median Unbiased Estimation of Coefficient Variance in a Time Varying Parameter Model," *Journal of the American Statistical Association* 93, 349-358.
- Stock, J.H. and M.W. Watson (2002a), "Has the Business Cycle Changed and Why?" forthcoming, *NBER Macroeconomics Annual* 2002.
- Stock, J.H. and M.W. Watson (2002b), "Forecasting Using Principal Components from a Large Number of Predictors," *Journal of the American Statistical Association* 97, 1167 – 1179.

Table A.1 Sources for GDP Data

Country	Series Name	Source	Full sample period
Canada	cngdp...d	CANSIM - STATISTICS CANADA (DataStream)	1961:1 2002:4
France	frona017g	GDP Vol. Index / OECD (DataStream)	1960:1 2002:4
Germany	bdgdp\$\$\$d	DEUTSCHE BUNDESBANK (DataStream)	1960:1 2002:4
Italy	iti99bvrg	IMF GDP Volume Index (DataStream)	1960:1 2001:4
Japan	jpona017g	GDP Vol. Index / OECD (DataStream)	1960:1 2002:4
UK	ukgdp\$\$\$d	OFFICE FOR NATIONAL STATISTICS (DataStream)	1960:1 2002:4
US	gdpq	citibase	1960:1 – 2002:3

Note: Data for earlier dates are available for some countries but were not used in this study. Four apparent outliers associated with strikes and German unification (Italy 1966:I, France 1968:II-1968:III and Germany 1991:I) were replaced with the series-specific full sample median growth rate.

Table 1. Mean and Standard Deviations of 4-quarter GDP growth in the G7 by decade

	Mean growth rate				Standard deviation of 4-quarter growth			
	60-69	70-79	80-89	90-02	60-69	70-79	80-89	90-02
Canada	3.81	2.54	1.71	1.52	1.50	1.82	2.67	2.24
France	4.07	2.93	1.78	1.32	1.23	1.66	1.24	1.40
Germany	3.83	2.59	1.57	1.31	2.56	2.13	1.67	1.53
Italy	4.45	3.17	2.09	1.34	2.44	3.28	1.58	1.28
Japan	8.65	3.55	3.21	1.21	2.19	3.16	1.57	2.08
UK	2.35	2.20	2.12	1.81	1.84	2.48	2.51	1.60
US	2.95	2.20	2.00	1.63	2.09	2.74	2.66	1.48

Notes: The columns labeled “Mean growth rate” show the sample mean of $400 \times \ln(GDP_t / GDP_{t-1})$. The “Standard deviation” columns show the standard deviation of $100 \times (GDP_t / GDP_{t-4})$.

Table 2. Test for Breaks in Autoregressive Parameters

	Conditional mean			Conditional variance: break model			Conditional variance: Trend + break model		
	<i>p</i>	Break date	67% confidence interval	<i>p</i>	Break date	67% confidence interval	<i>p</i> : trend	<i>p</i> : break	Break date
Canada	0.05			0.00	1991:2	1990:4 - 1993:1	0.00	0.17	.
France	0.00	1974:1	1973:3 - 1974:3	0.01	1968:1	1967:3 - 1970:2	0.84	0.60	.
Germany	0.39			0.00	1993:1	1992:3 - 1995:2	0.49	0.91	.
Italy	0.02	1990:1	1989:3 - 1990:3	0.01	1981:3	1980:3 - 1984:4	0.95	0.42	.
Japan	0.00	1973:1	1972:3 - 1973:3	0.32			0.34	0.11	.
UK	0.00	1980:1	1979:3 - 1980:3	0.00	1980:1	1979:4 - 1982:1	0.00	0.00	1970:4
US	0.99			0.00	1983:2	1982:4 - 1985:3	0.65	0.00	1983:2

**Table 3. Autoregressive Parameters for GDP growth rates:
Sums of AR coefficients and standard error of the regression**

$$\Delta y_t = \mu + \alpha(L)\Delta y_{t-1} + \varepsilon_t$$

	$\hat{\alpha}(1)$		$\hat{\sigma}_\varepsilon$	
	1960 - 1983	1984 - 2002	1960 - 1983	1984 - 2002
Canada	0.18	0.55	3.81	2.29
France	-0.42	0.40	2.87	1.64
Germany	0.04	-0.25	5.42	3.43
Italy	0.20	0.14	3.73	2.22
JP	0.17	-0.10	4.07	3.76
UK	0.00	0.68	4.80	1.85
US	0.32	0.44	4.05	2.05

Notes: Estimates are computed using detrended GDP growth.

Table 4. Cross-country GDP growth correlations

(a) Four-quarter growth rates, simple correlation coefficients

	Canada	France	Germany	Italy	Japan	UK	US
1960 - 1983							
Canada	1.00
France	0.29	1.00
Germany	0.55	0.57	1.00
Italy	0.39	0.54	0.37	1.00	.	.	.
Japan	0.22	0.37	0.47	0.20	1.00	.	.
UK	0.28	0.54	0.53	0.18	0.50	1.00	.
US	0.76	0.37	0.56	0.35	0.34	0.49	1.00
1984 - 2002							
Canada	1.00
France	0.38	1.00
Germany	0.15	0.58	1.00
Italy	0.42	0.76	0.54	1.00	.	.	.
Japan	-0.04	0.26	0.33	0.30	1.00	.	.
UK	0.74	0.32	0.08	0.48	0.07	1.00	.
US	0.80	0.28	0.19	0.28	-0.01	0.58	1.00

(b) Four-quarter growth rates, implied by reduced form VAR(4,1)

	Canada	France	Germany	Italy	Japan	UK	US
1960 - 1983							
Canada	1.00
France	0.27	1.00
Germany	0.58	0.54	1.00
Italy	0.37	0.51	0.41	1.00	.	.	.
Japan	0.29	0.26	0.39	0.16	1.00	.	.
UK	0.36	0.53	0.51	0.18	0.48	1.00	.
US	0.73	0.34	0.54	0.30	0.26	0.46	1.00
1984 - 2002							
Canada	1.00
France	0.60	1.00
Germany	0.13	0.56	1.00
Italy	0.52	0.81	0.47	1.00	.	.	.
Japan	-0.02	0.13	0.31	0.12	1.00	.	.
UK	0.72	0.59	0.20	0.62	0.03	1.00	.
US	0.82	0.69	0.27	0.49	0.04	0.69	1.00

Table 4, continued

(c) BP-filtered GDP, implied by reduced form VAR(4,1)

	Canada	France	Germany	Italy	Japan	UK	US
	1960 – 1983						
Canada	1.00
France	0.32	1.00
Germany	0.64	0.58	1.00
Italy	0.41	0.56	0.45	1.00	.	.	.
Japan	0.31	0.28	0.42	0.17	1.00	.	.
UK	0.40	0.58	0.56	0.22	0.52	1.00	.
US	0.76	0.41	0.60	0.34	0.28	0.51	1.00
	1984 – 2002						
Canada	1.00
France	0.47	1.00
Germany	0.09	0.62	1.00
Italy	0.45	0.80	0.50	1.00	.	.	.
Japan	-0.03	0.19	0.40	0.18	1.00	.	.
UK	0.67	0.47	0.15	0.58	0.05	1.00	.
US	0.77	0.60	0.25	0.40	0.07	0.62	1.00

Notes: The correlations implied by the reduced form VAR(4,1) were computed from the VAR coefficients and variance matrix estimated over the two subsamples. The correlations for BP-filtered GDP were computed for an idealized BP filter.

Table 5. Tests of k -factor Model versus Unrestricted Error Covariance Matrix

Number of Factors (k)	d.f.	1960-2002		1960-1983		1964-2002	
		LR Statistic	p -value	LR Statistic	p -value	LR Statistic	p -value
1	14	40.46	0.00	29.10	0.01	33.30	.00
2	8	9.70	0.29	9.89	0.27	8.60	.38
3	3	1.98	0.58	0.48	0.92	0.87	0.83

Notes: Entries are the likelihood ratio test statistic and its p -value testing the null hypothesis that the VAR(4,1) error covariance matrix has a k -factor structure, against the unrestricted alternative. The degrees of freedom of the test are given in the second column.

Table 6. Variance Decompositions: Common Shocks, Spillovers, and Own-Country Shocks

(a) Forecast Error Variance Decompositions for GDP Growth

	horizon	1960 – 1983				1984 - 2002			
		standard deviation	Factor	Spillover	Own	standard deviation	Factor	Spillover	Own
Canada	1	3.39	0.42	0.00	0.58	2.08	0.95	0.00	0.05
	2	2.75	0.50	0.08	0.41	1.77	0.92	0.04	0.04
	4	2.08	0.54	0.15	0.31	1.72	0.89	0.08	0.03
	8	1.54	0.55	0.16	0.28	1.67	0.84	0.13	0.03
France	1	2.68	0.88	0.00	0.12	1.47	0.99	0.00	0.01
	2	1.87	0.82	0.09	0.09	1.15	0.95	0.04	0.01
	4	1.25	0.77	0.16	0.07	1.04	0.93	0.07	0.01
	8	0.84	0.76	0.18	0.06	1.01	0.89	0.11	0.01
Germany	1	4.90	0.30	0.00	0.70	3.14	0.25	0.00	0.75
	2	3.35	0.39	0.08	0.53	2.04	0.31	0.06	0.63
	4	2.30	0.44	0.14	0.43	1.25	0.34	0.06	0.59
	8	1.75	0.47	0.15	0.38	0.89	0.40	0.07	0.53
Italy	1	3.57	0.08	0.00	0.92	1.96	0.37	0.00	0.63
	2	3.17	0.09	0.01	0.90	1.39	0.43	0.07	0.50
	4	2.66	0.11	0.03	0.85	1.11	0.47	0.10	0.43
	8	1.82	0.13	0.04	0.82	0.92	0.54	0.14	0.32
Japan	1	3.78	0.17	0.00	0.83	3.55	0.01	0.00	0.99
	2	2.83	0.20	0.02	0.78	2.41	0.01	0.03	0.96
	4	2.26	0.20	0.03	0.77	1.68	0.01	0.04	0.96
	8	1.62	0.21	0.03	0.76	1.21	0.00	0.05	0.95
U.K.	1	4.69	0.27	0.00	0.73	1.68	0.03	0.00	0.97
	2	3.20	0.26	0.04	0.70	1.56	0.10	0.01	0.90
	4	2.36	0.28	0.05	0.67	1.31	0.20	0.02	0.78
	8	1.70	0.29	0.06	0.65	1.23	0.30	0.04	0.66
U.S.	1	3.99	0.26	0.00	0.74	1.73	0.21	0.00	0.79
	2	3.23	0.30	0.01	0.69	1.40	0.29	0.05	0.66
	4	2.53	0.32	0.02	0.66	1.29	0.37	0.12	0.51
	8	1.87	0.33	0.03	0.64	1.23	0.47	0.18	0.36

Table 6, continued**(b) BP-filtered GDP**

	1960 – 1983				1984 – 2002			
	standard deviation	Factor	Spillover	Own	standard deviation	Factor	Spillover	Own
Canada	1.23	0.54	0.19	0.27	1.14	0.81	0.17	0.03
France	0.72	0.72	0.22	0.06	0.73	0.85	0.14	0.00
Germany	1.34	0.46	0.18	0.36	0.72	0.41	0.09	0.50
Italy	1.65	0.13	0.05	0.82	0.69	0.51	0.14	0.34
Japan	1.34	0.21	0.03	0.76	0.95	0.02	0.05	0.93
U.K.	1.35	0.29	0.07	0.64	0.84	0.33	0.05	0.62
U.S.	1.50	0.33	0.03	0.63	0.85	0.43	0.20	0.37

Notes: Standard deviations are in percent per year (growth rates) or percentages (BP-filtered). Remaining entries are the fraction of the variance associated with the common factors, with cross-country spillovers, or with the country's own idiosyncratic shocks. Panel (a) presents forecast error variance decompositions at various horizons; panel (b) decomposes the total variance of BP-filtered GDP.

Table 7. Decomposition of changes in the variance of 4-quarter-ahead forecast errors into changing impulses and changing propagation.

	Variances			Contribution of change in shock variance				Contribution of change in impulse response function			
	1960-1983	1984-2002	change	int'l	spill-over	own	total	int'l	spill-over	own	total
Canada	4.31	2.95	-1.37	-7.86	-0.64	-2.23	-10.73	8.16	0.23	0.97	9.37
France	1.57	1.09	-0.48	-2.31	-0.29	-0.19	-2.78	2.11	0.11	0.08	2.30
Germany	5.31	1.56	-3.75	-1.68	-0.40	-1.23	-3.31	-0.10	-0.22	-0.12	-0.44
Italy	7.07	1.23	-5.84	-1.46	-0.43	-3.44	-5.33	1.22	0.33	-2.07	-0.51
Japan	5.09	2.82	-2.27	-0.46	-0.23	0.19	-0.50	-0.56	0.21	-1.42	-1.77
U.K.	5.56	1.72	-3.84	-1.57	-0.25	-4.78	-6.59	0.34	0.01	2.40	2.75
U.S.	6.42	1.66	-4.76	-2.55	-0.55	-3.38	-6.48	1.11	0.61	0.00	1.72

Notes: The first three columns give the variance of BP-filtered GDP in the first and second subsample, estimated using the FSVAR, and their difference. The remaining columns decompose this difference into changes in the impulse response functions and changes in the variances of the shocks themselves. The sum of the “international,” “spillover,” and “own” columns equals the “total” column, and the sum of the two “total” columns equals the “change” column.

Table 8. Counterfactual Volatility Measures during 1984 – 2002 using Common and Country Shock Variances from 1960 – 1983

(a) Standard Deviations of 4-quarter GDP growth

Period for shock variances		Standard deviation of 4-quarter GDP growth						
Common shocks	Country shocks	Canada	France	Germany	Italy	Japan	UK	US
84-02	84-02	2.20	1.43	1.35	1.30	1.70	1.63	1.68
60-83	84-02	5.03	3.08	2.06	2.36	1.78	2.76	3.16
60-83	60-83	5.71	3.43	2.46	3.05	1.92	4.03	4.10

(b) Standard Deviations of BP-filtered GDP

Period for shock variances		Standard deviation of BP-filtered GDP						
Common shocks	Country shocks	Canada	France	Germany	Italy	Japan	UK	US
84-02	84-02	1.14	0.73	0.72	0.69	0.95	0.84	0.85
60-83	84-02	2.61	1.53	1.09	1.24	0.99	1.40	1.55
60-83	60-83	2.95	1.69	1.29	1.61	1.06	2.09	2.06

Notes: Entries in panel (a) are the standard deviations of four-quarter GDP growth based on the FSVAR impulse response functions estimated using data from 1984 – 2002, using the shock variances estimated over the sample indicated in the first two columns. The first row is the model-based estimate of the actual standard deviation during 1984 – 2002; the remaining rows are counterfactuals. The entries in panel (b) are analogous to those in panel (a) but pertain to BP-filtered GDP.

Table 9. Counterfactual Correlations Between Four-quarter Growth Rates during 1984 – 2002 using Common Shock Variances from 1960 – 1983

(a) FSVAR estimates of actual 1984 – 2002 correlations

	Canada	France	Germany	Italy	Japan	UK	US
Canada	1.00
France	0.61	1.00
Germany	0.14	0.57	1.00
Italy	0.54	0.82	0.49	1.00	.	.	.
Japan	0.00	0.13	0.24	0.10	1.00	.	.
UK	0.73	0.59	0.21	0.60	0.06	1.00	.
US	0.81	0.70	0.26	0.56	0.12	0.72	1.00

(b) FSVAR estimates of 1984 – 2002 correlations using common shock variances from 1960 - 1983

	Canada	France	Germany	Italy	Japan	UK	US
Canada	1.00
France	0.72	1.00
Germany	0.30	0.73	1.00
Italy	0.73	0.93	0.70	1.00	.	.	.
Japan	0.07	0.17	0.27	0.15	1.00	.	.
UK	0.87	0.75	0.44	0.78	0.16	1.00	.
US	0.91	0.85	0.48	0.81	0.17	0.89	1.00

Notes: Entries in panel (a) are the correlations among four-quarter GDP growth based on the FSVAR estimated using data from 1984 – 2002. Entries in panel (b) are based on the 1984 – 2002 FSVAR, except calculated using the common shock variances from the 1960 – 1983 FSVAR.

Table 10. Counterfactual Standard Deviation of 4-quarter GDP Growth Based on Trade-Weighted FSVAR

Period for shock variances		Standard deviation of 4-quarter GDP growth						
Common shocks	Country shocks	Canada	France	Germany	Italy	Japan	UK	US
84-02	84-02	1.96	1.16	1.34	1.16	1.76	1.50	1.46
60-83	84-02	4.63	2.36	1.87	1.89	1.89	1.77	2.58
60-83	60-83	5.11	2.48	2.27	2.50	2.00	3.60	3.44

Notes: Entries are computed in the same way as in panel (a) of Table 8, except they are based on the FSVAR (7) with trade-weight lag restrictions.

Table 11. Sensitivity Check: Counterfactual Correlations of 4-quarter GDP Growth Based on Trade-Weighted FSVAR

(a) Trade-weighted FSVAR estimates of actual 1984 – 2002 correlations

	Canada	France	Germany	Italy	Japan	UK	US
Canada	1.00
France	0.21	1.00
Germany	-0.08	0.51	1.00
Italy	0.17	0.70	0.40	1.00	.	.	.
Japan	0.17	0.12	0.12	0.08	1.00	.	.
UK	0.17	0.35	0.21	0.24	0.08	1.00	.
US	0.74	0.35	0.11	0.27	0.28	0.20	1.00

(b) Trade-weighted FSVAR estimates of 1984 – 2002 correlations using common shock variances from 1960 - 1983

	Canada	France	Germany	Italy	Japan	UK	US
Canada	1.00
France	0.40	1.00
Germany	-0.08	0.60	1.00
Italy	0.38	0.87	0.53	1.00	.	.	.
Japan	0.25	0.20	0.19	0.14	1.00	.	.
UK	0.39	0.57	0.32	0.50	0.18	1.00	.
US	0.90	0.55	0.11	0.52	0.33	0.46	1.00

Notes: Entries are computed in the same way as in Table 9, except they are based on the FSVAR (7) with trade-weight lag restrictions.

Table 12. Sensitivity Check: Counterfactual Square Roots of Average Coherences over Business Cycle Frequencies

(a) FSVAR estimated actuals for 1984 – 2002

	Canada	France	Germany	Italy	Japan	UK	US
Canada	1.00
France	0.78	1.00
Germany	0.39	0.64	1.00
Italy	0.72	0.91	0.60	1.00	.	.	.
Japan	0.17	0.15	0.31	0.16	1.00	.	.
UK	0.85	0.75	0.38	0.73	0.18	1.00	.
US	0.90	0.82	0.42	0.73	0.18	0.85	1.00

(b) FSVAR estimated counterfactuals for 1984 – 2002 using common shock variances from 1960 - 1983

	Canada	France	Germany	Italy	Japan	UK	US
Canada	1.00
France	0.85	1.00
Germany	0.60	0.83	1.00
Italy	0.86	0.97	0.81	1.00	.	.	.
Japan	0.28	0.21	0.31	0.24	1.00	.	.
UK	0.95	0.84	0.60	0.86	0.28	1.00	.
US	0.96	0.92	0.66	0.91	0.27	0.95	1.00

Notes: Entries are square root of the average coherence at business cycle frequencies, as defined in (8), computed using the FSVAR described in Section 4. The factual (panel (a)) and counterfactual (panel (b)) scenarios are the same as in Table 9.

Figure 1. Four-quarter growth rates of GDP

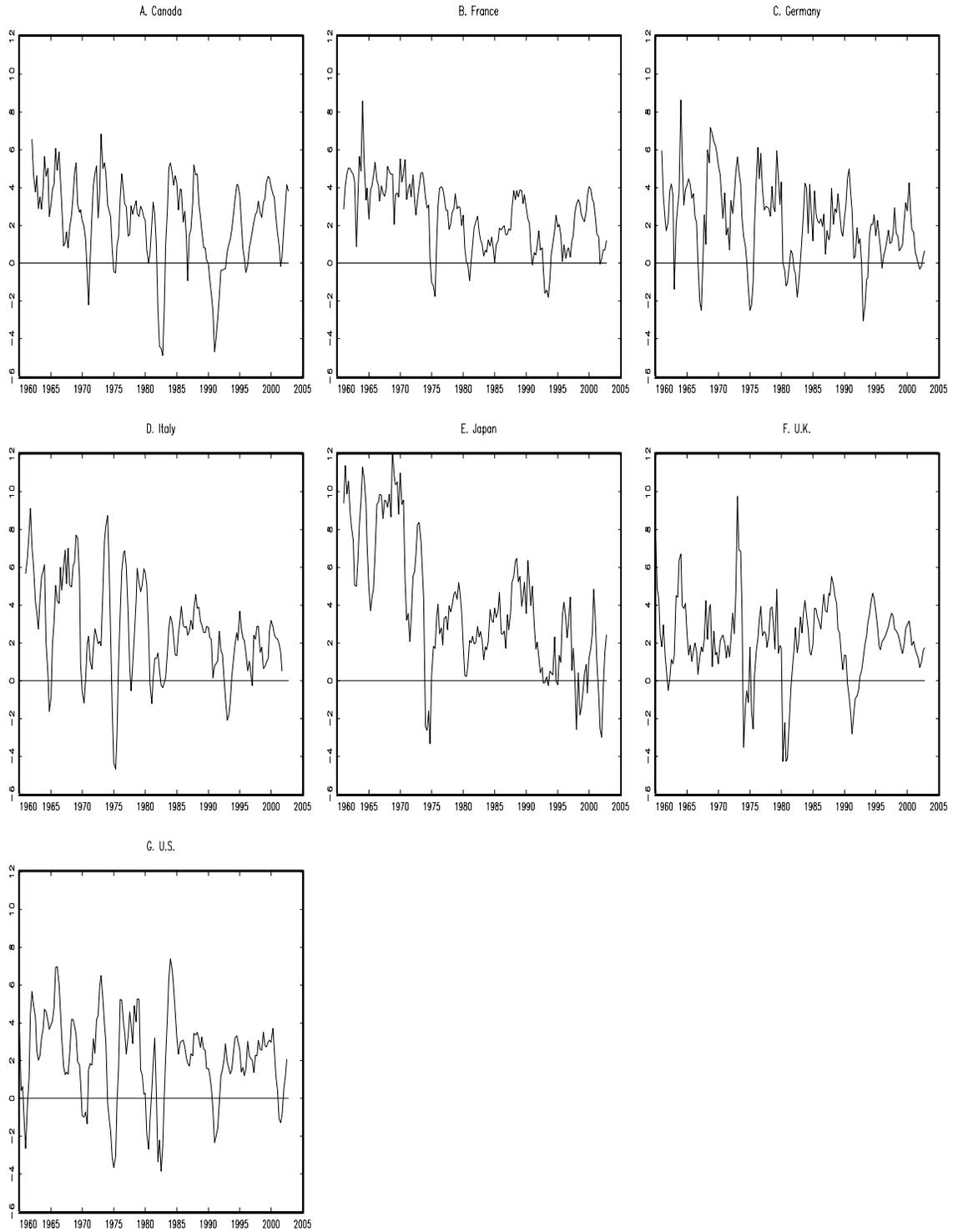


Figure 2. Detrended four-quarter GDP growth:
individual countries (solid lines) and G7 aggregate (dashed line)

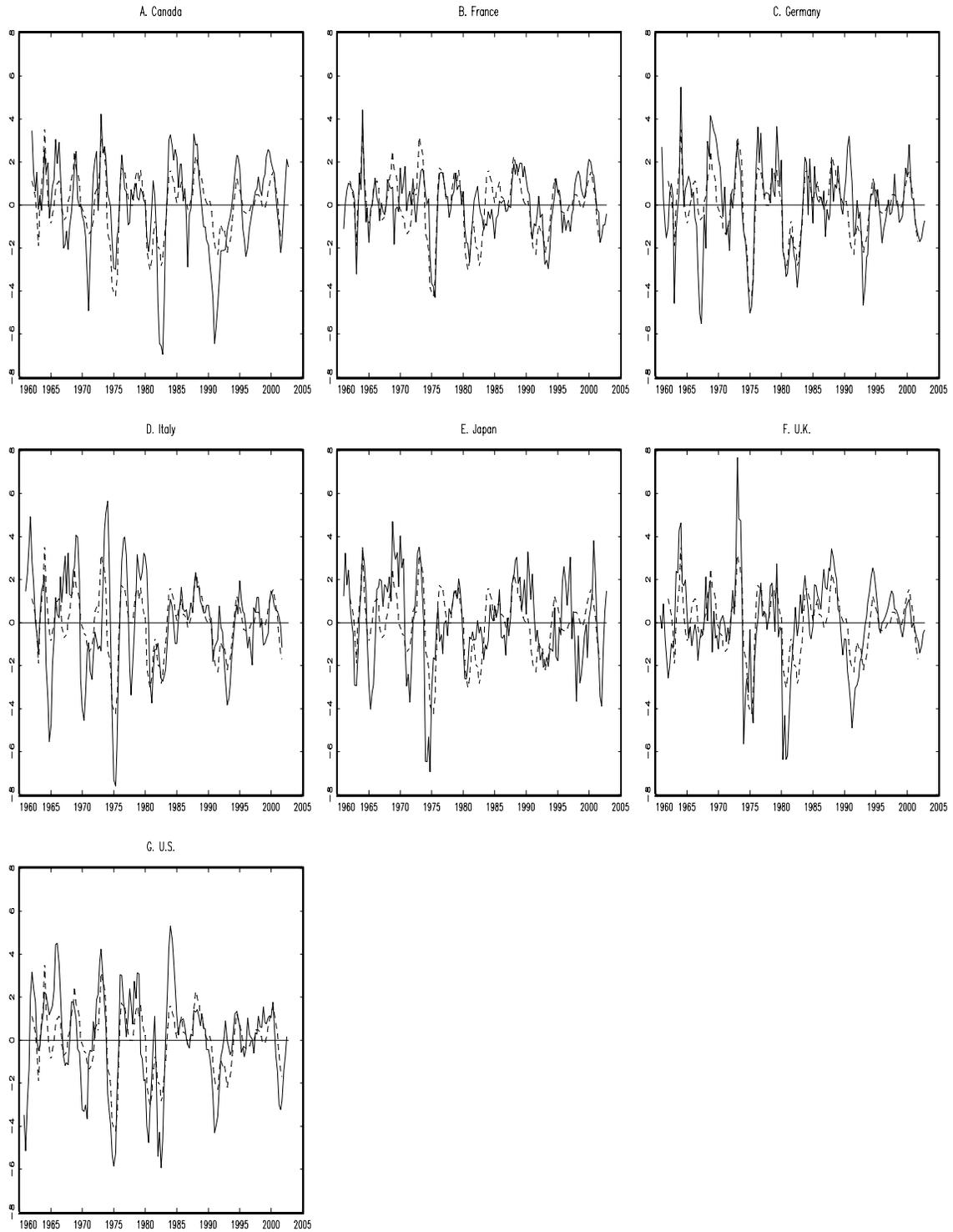


Figure 3. Band-pass GDP growth:
individual countries (solid lines) and G7 aggregate (dashed line)

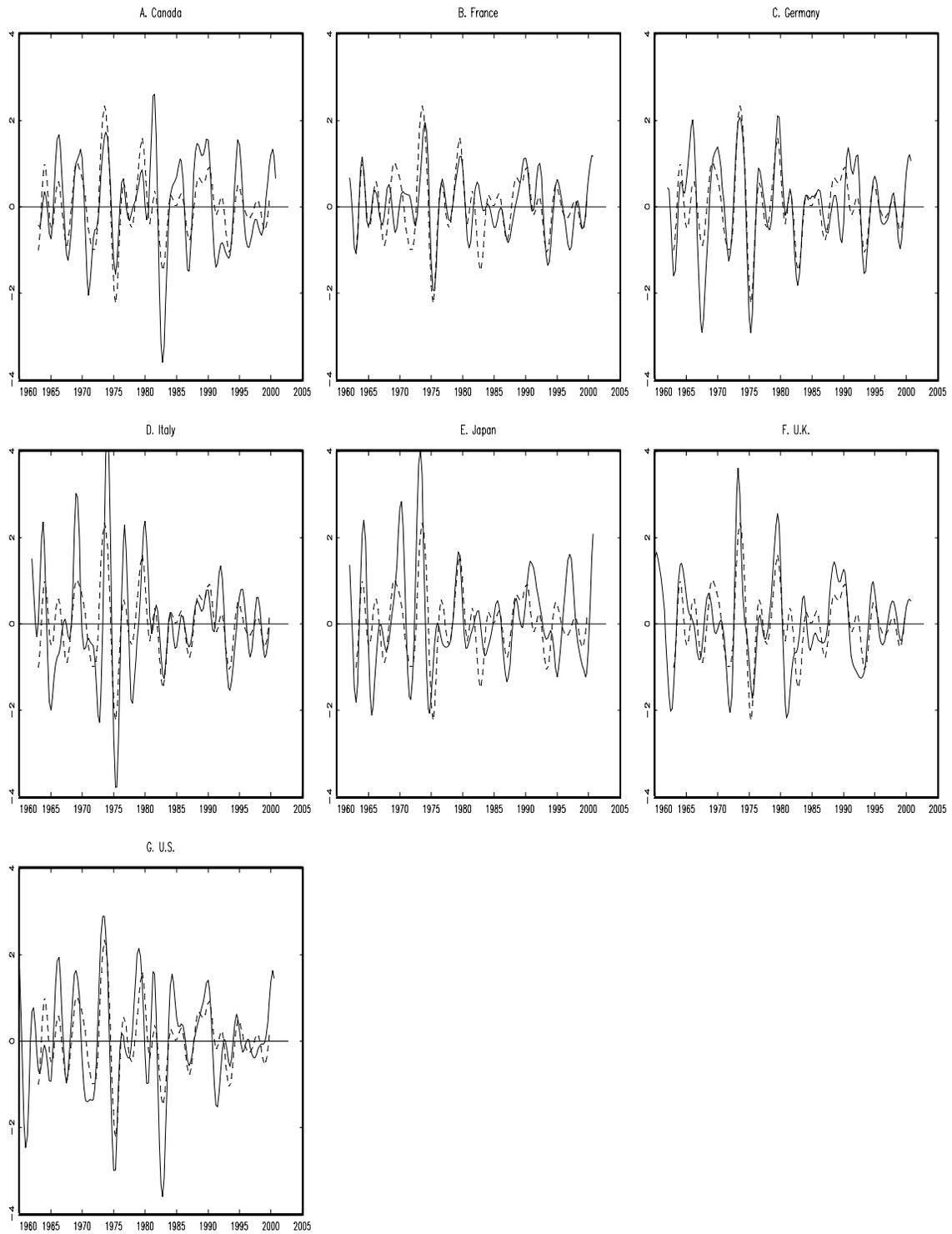


Figure 4. Estimated instantaneous standard deviation of 4-quarter GDP growth

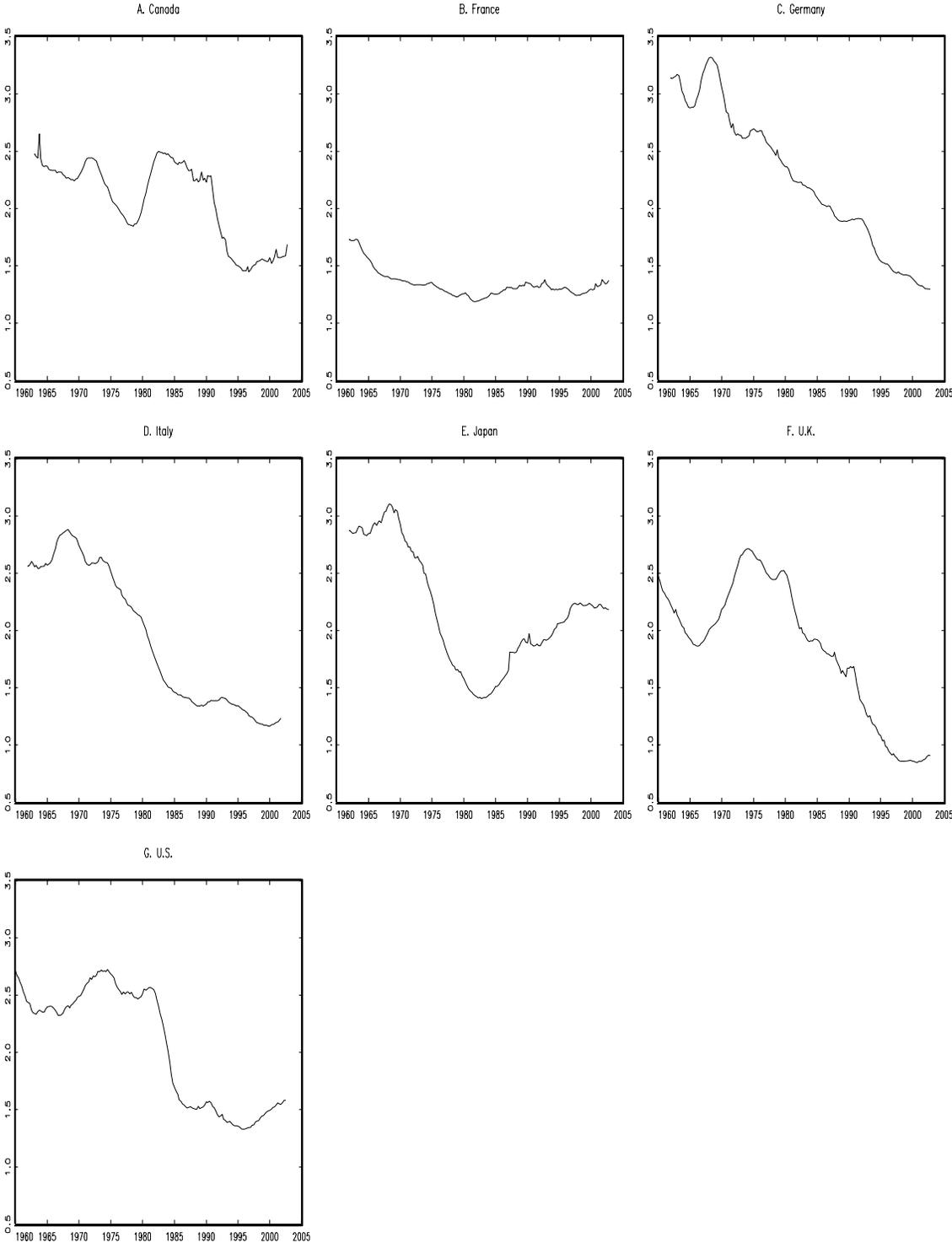


Figure 5. Estimated instantaneous standard deviation of BP GDP

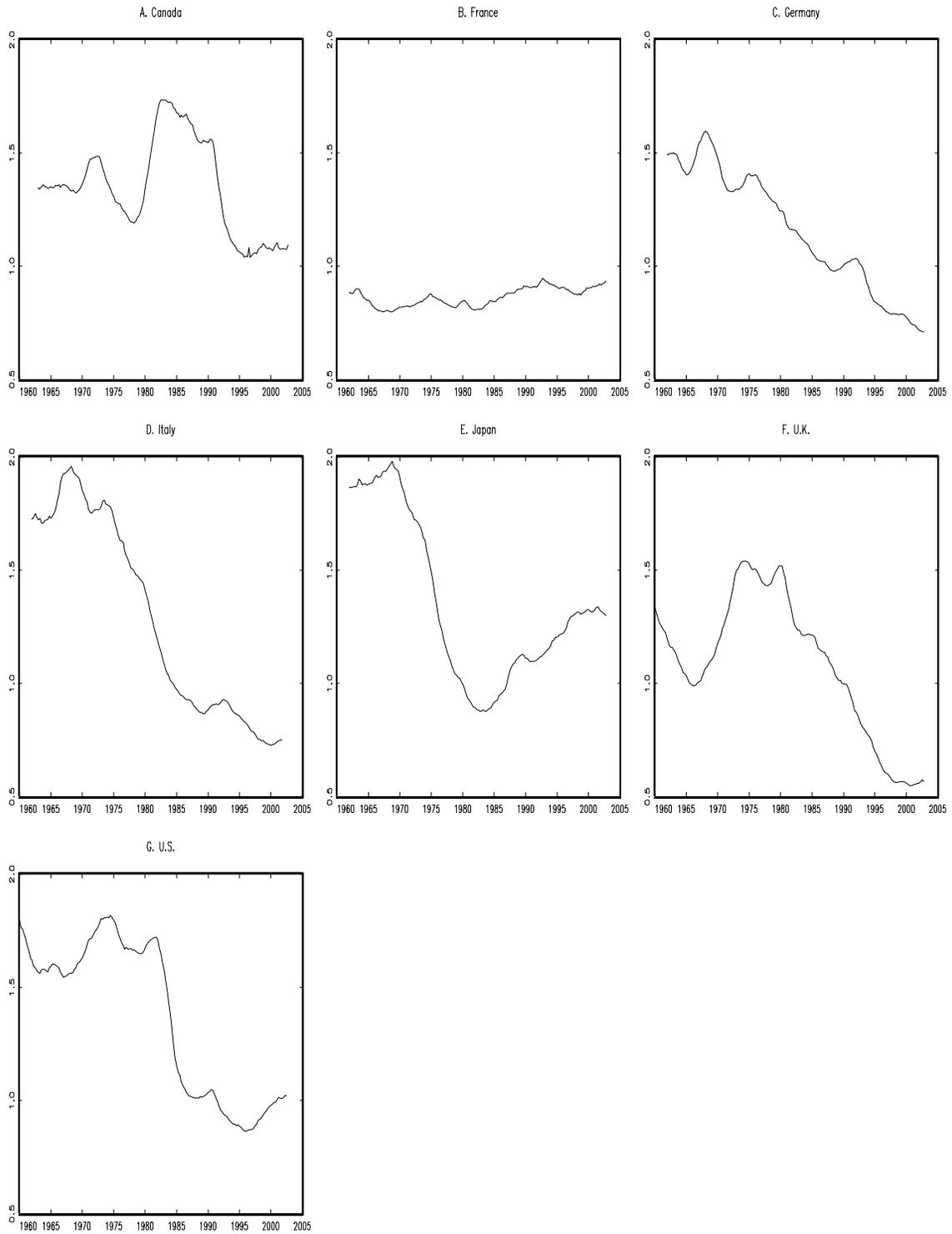


Figure 6a. Rolling autoregressions: sum of AR coefficients ($\hat{\alpha}(1)$)

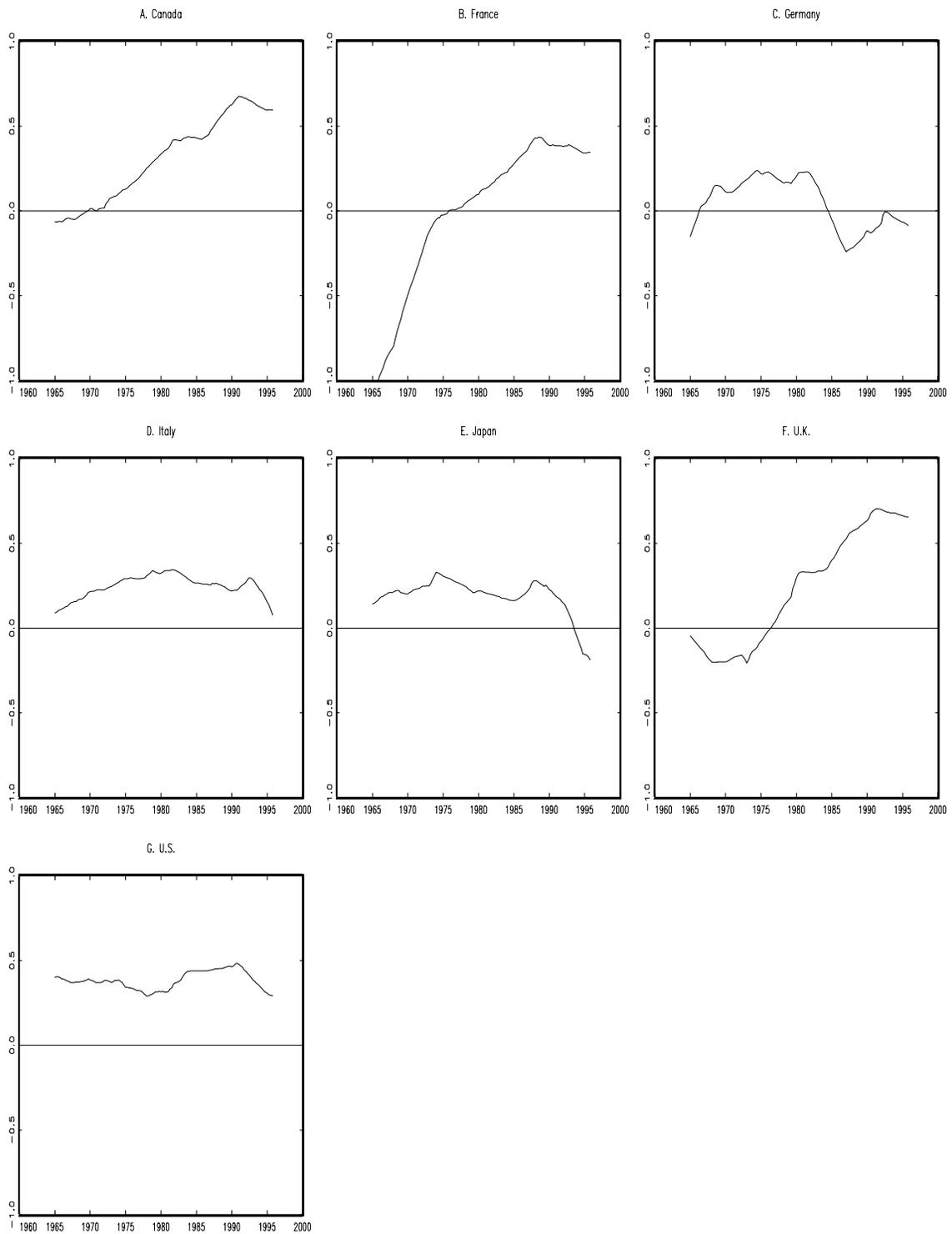


Figure 6b. Rolling autoregressions: innovation standard error ($\hat{\sigma}_\varepsilon$)

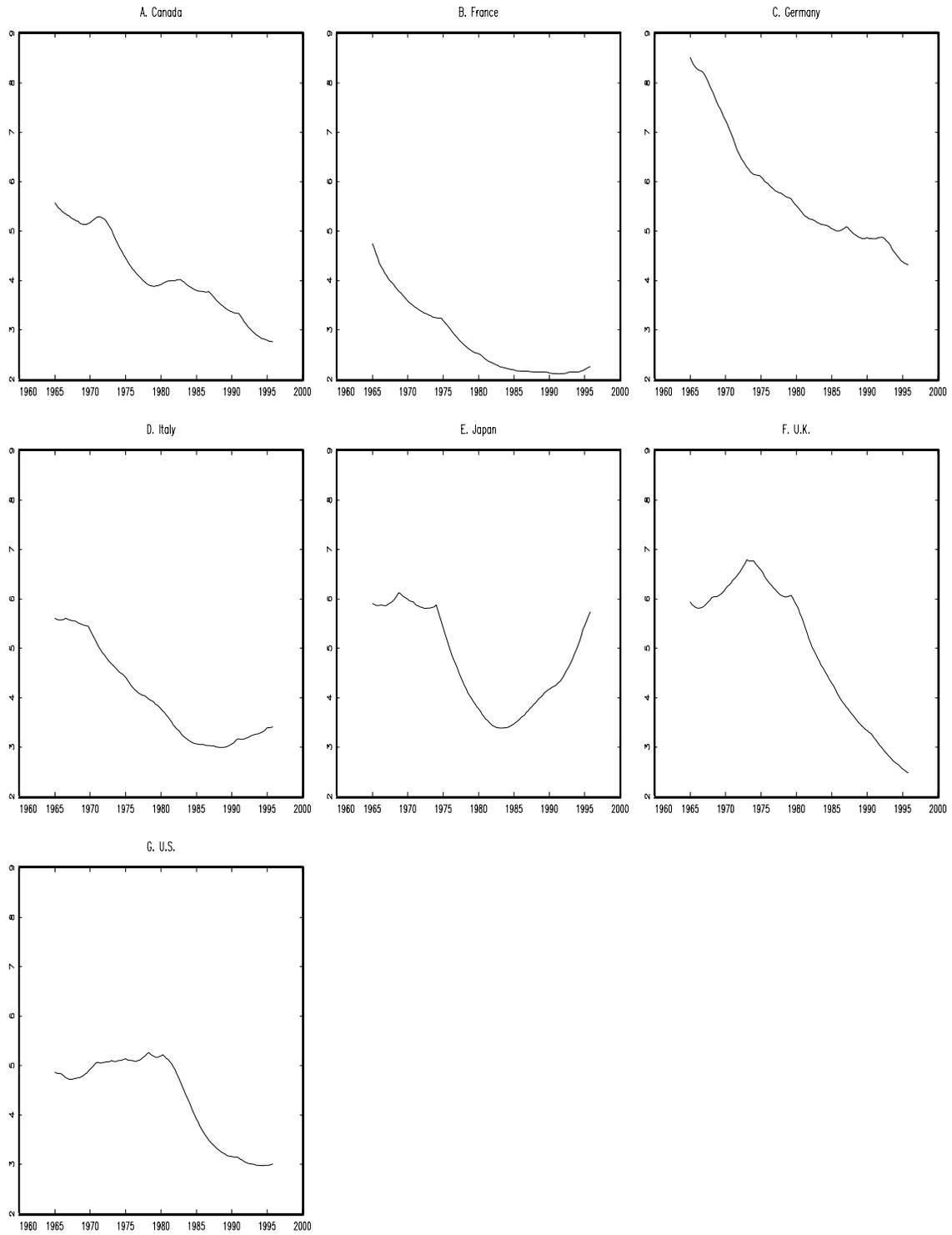


Figure 7. Band-pass GDP growth:
rolling correlation with US (solid line) and Germany (dashed line)

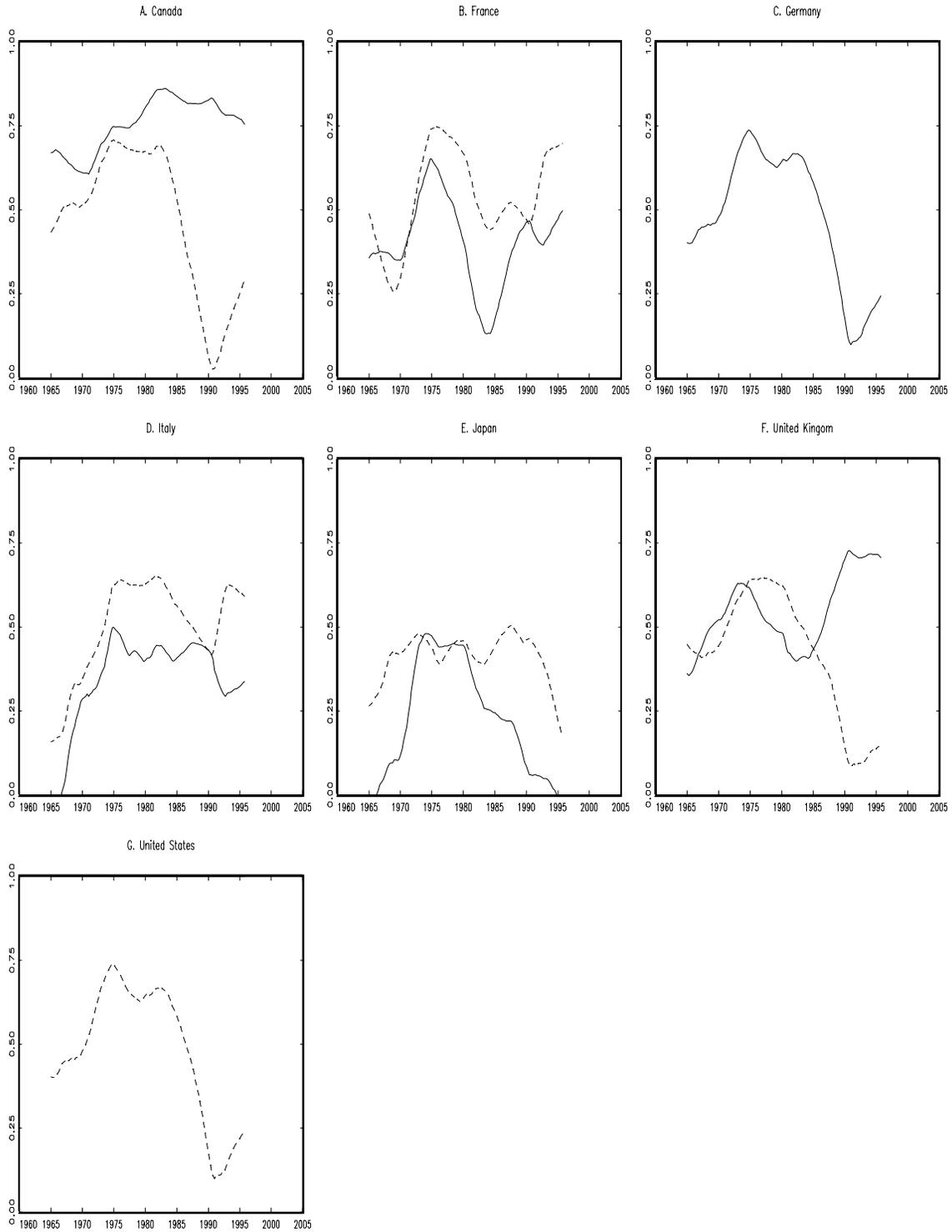


Figure 8. Rolling FSVAR Variance decomposition of BP-filtered GDP growth: International shocks (lower); international shocks + spillovers (middle); and total (top)

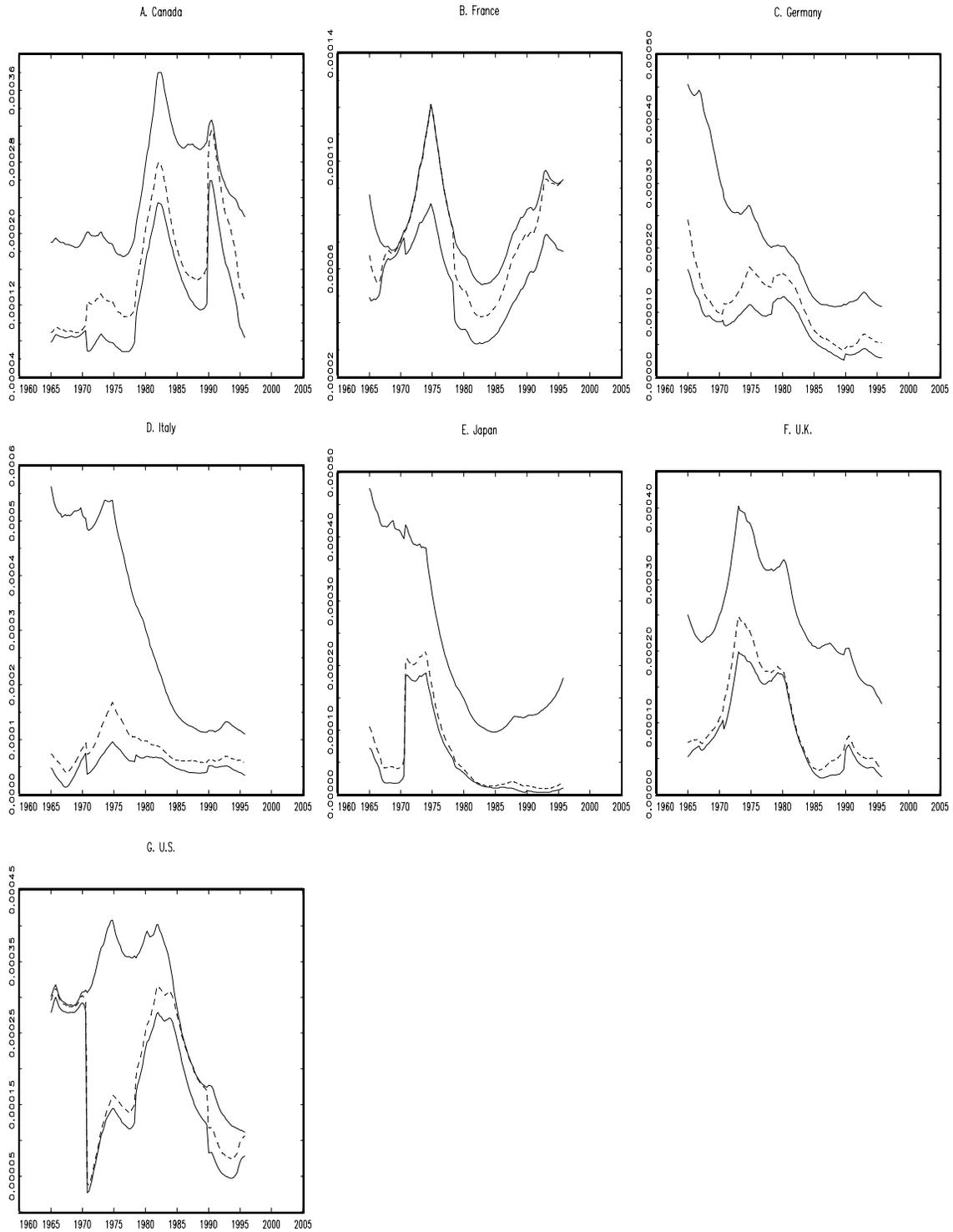


Figure 9a. Cumulative impulse response function: first common factor
in 1960 – 198 (solid line) and 1984 – 2002 (dashed line)

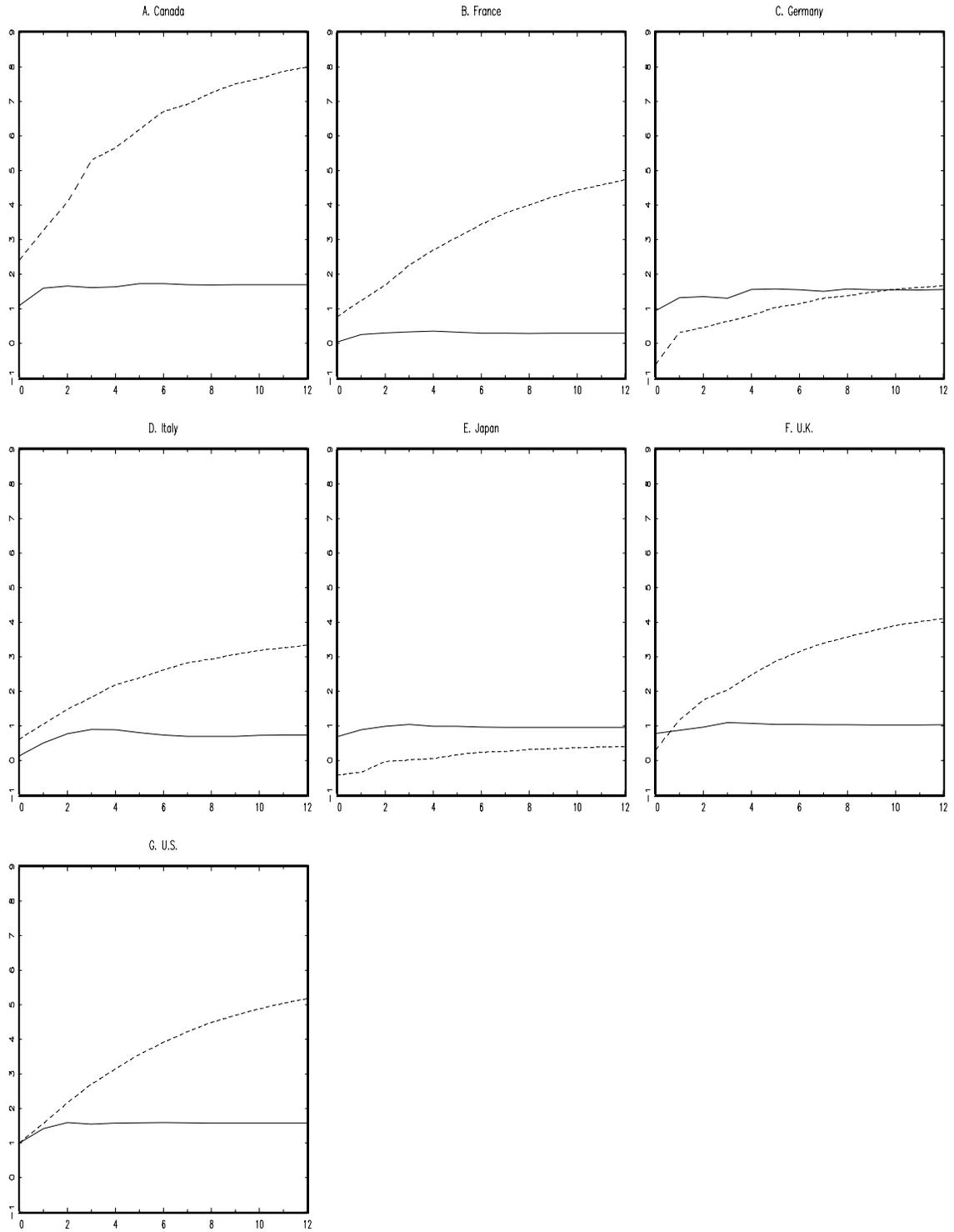


Figure 9b. Cumulative impulse response function: second common factor in 1960 – 198 (solid line) and 1984 – 2002 (dashed line)

