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We revisit the macroeconomic effects of labor reallocation within the framework of Campbell and Kuttner (1996). We re-estimate their model, update the sample, and employ generalized and local impulse response functions. We confirm that total employment responses to reallocation shocks remain significant.

JEL classification: C32, C50, E24

Keywords: labor reallocation, unemployment, sectoral shifts

1 Introduction

Campbell and Kuttner 1996 (CK hereafter) provide a seminal paradigm for testing and measuring the macroeconomic effects of labor reallocation devoid of dispersion measures à la Lilien (1982) and so bypass difficulties linked to such proxies (see Gallipoli and Pelloni, 2013, for a review of the literature).

This paper extends CK in four directions: first, we extend the sample (60 years overall); second, we consider recent developments in time series econometrics (Jorda 2005 and 2009); third, we take into account the intrinsic asymmetries of sectoral shifts following Kilian and Vigfusson (2011); and fourth, we consider the effects of real exchange rate shocks on reallocations. This analysis focuses on the response of total employment to sectoral shocks (impulse responses within a VAR framework).

The results confirm CK that sectoral shifts play a significant role in the fluctuations of total employment. Moreover, there is evidence that the magnitude of the impact of reallocation shocks has decreased when one considers an up to date sample.

2 Sectoral Shifts

We estimate a twelve lags specification of CK's bivariate SVAR. The reduced form of the model for ρ lags can be written as follow:

$$y_{t} = By_{t} + A_{1}y_{t-1} + A_{2}y_{t-2} + \dots + A_{\rho}y_{t-\rho} + \varepsilon_{t}$$

where $y_t = [DN, DW]$ is the vector of the endogenous variables, *B* is an identity 2x2 matrix for the contemporaneous effects, A_{ρ} is a 2x2 coefficient matrix for the lagged values and $\varepsilon_t = [u_t, v_t]$ is the vector that contains the uncorrelated aggregate and sectoral shocks respectively. The endogenous variables are log-differences of total employment, DN, and manufacturing's share, DW (following CK's notation). We employ short-run and long-run restrictions (see the appendix).

Real exchange rate changes could affect unemployment via both an aggregate and a reallocation channel. However, Gourinchas (1998), Davis and Haltiwanger (2001), Klein et al. (2003) and Haltiwanger et al. (2004) bear out that both job creation and destruction would move in the same direction in response to a real exchange rate disturbance. Such a response should support the view of a pure reallocation shock, since an aggregate shock should have brought about diverging responses of job creation and destruction. Thus we add a third equation for the real effective exchange rate (E) to the baseline bivariate VAR and impose the same restrictions as in the case of DZ.

We also re-estimate CK's seven-dimensional VAR (employment shares of six sectors and aggregate employment) to capture much of the inter-sectoral reallocation missed by the bivariate structure. We adopt the same sectoral disaggregation of CK (construction; fire, insurance and real estate (FIRE); transportation; wholesale and retail trade; and Government) and do not orthogonalize sectoral shocks (no natural ordering of these variables).

We employ the generalised impulse responses (GIRFs) of Pesaran and Smith (1998) and the local impulse responses (local IRF) of Jorda (2005, 2009).

Finally, given the intrinsic asymmetries of sectoral shifts (Panagiotidis and Pelloni, 2014), we test the asymmetric responses of total employment to changes in manufacturing share. Following Kilian and Vigfusson (2011), we compute a Wald-type test¹ to examine the case of asymmetry. The VAR employed contains the vector of endogenous variables $y_t = [DN, DW^+, DW^-]$, where + and - indicate that manufacturing share is censored to contain only positives or only negative values².

3 Empirical Results

Table 1 presents the data. The sample ranges from 1955M02 to 2016M08. Table 2 summarises the specifications and reports the variance decomposition analysis of all the models together with the initial CK estimates. The focus is on the share of sectoral shocks to the variance of total employment. It is evident that the identification assumptions have an effect on the results. When sectoral shifts are restricted not to have a contemporaneous impact on aggregate variables, they account for a relatively smaller portion of the variance. When total employment does not affect the other sector in the long-run, the variance share of the sectoral shocks ranges from 48.1% (2 sectors) to 84.7% (7 sectors). Full sample results suggest that the impact of reallocation shocks tends to fall. When sector disaggregation increases, sectoral shocks account for 1/4 of the variance of total employment even within the unfavorable short-run restricted SVAR.

Figure 1 presents the impulse responses of the two-sectors model, comparing the CK's sample and the updated one. Shocks on the manufacturing sector affect total employment through the alternative functions and identifying assumptions. GIRFs and local IRFs from the simple VAR indicate for the CK's span that a one standard deviation shock to the growth rate of manufacturing's employment share causes an almost 0.1% increase in the growth rate of total employment. In comparison, sectoral shifts within the full sample affects aggregate employment for about 25% less. Shocks from the GIRFs are significant for almost six periods, while those from the local IRFs for two periods. Both produce positive responses. For the SVAR, the long-run restricted model behaves in line with the GIRFs and the local IRFs. The latter confirms the robustness of the CK's results to alternative methods of calculating the impulse responses.

¹The p-value of the relevant F-statistic is 0.0206. All results are available upon request.

²Note that sectoral shifts are not directional and only the size of the shock would matter. It is worth investigating whether the shock is significant in the case where this directional asymetry is taken into account.

Figure 2 refers to the model that introduces oil shocks that affect manufacturing contemporaneously and aggregate employment indirectly. The magnitude of the sectoral shocks identified with this, decreases by almost 50% for the GIRFs and 40% for the local IRFs (full sample). Shocks are statistically significant for six and two periods respectively. Innovations from SVARs have no statistical significance³.

In Figure 3, we replace total employment with total unemployment as well as oil price shocks with commodity price and exchange rate disturbances. Structural shocks continue to have no impact. Both commodity price and exchange rate innovations influence unemployment, through the manufacturing sector, in the same manner. The responses are negative and significant for three periods.

Finally, Figure 4 shows the asymmetric impulse response analysis in the sense of Killian and Vigfusson (2011) for the full sample. From both the GIRF and the local IRF the impact of the shock coming from negative growth rates of the manufacturing share seems to be greater than the positive ones, in terms of their mean point estimates⁴.

4 Conclusions

This paper examines the importance of labor reallocation. Alternative impulse response functions and variance decomposition analysis for VARs and SVARs are estimated to gauge the impact of sectoral shocks. Results from the benchmark model (CK) indicate that sectoral shifts' innovations are reduced by 25% when incorporating the up-to-date sample. Although the magnitude of this impact has decreased, responses of total employment to reallocation shocks remain statistically significant. The alternative methods for calculating the impulse responses provide similar conclusions, revealing that the CK results are robust. In the case of the 7 sector disaggregation, the variance decomposition analysis shows that, even in the most unfavorable scenario for sectoral shocks specification, these shocks account for almost 20% of the variance of aggregate employment. Reallocation shocks remain a significant source of aggregate employment's fluctuations.

³Thus, we cannot refer to manufacturing as a transmission channel.

⁴This difference is close to 15%. Nonetheless, the impact response for the negative sectoral values appears to be within the confidence interval of the positive ones and vice versa.

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	Description	Unemployment level, Thousands of Persons	Total Nonfarm Payrolls, Thousands of Persons	Producer Price Index for All Commodities, Index 1982=100	Real Narrow Effective Exchange Rate, Index 2010=100	Crude Petroleum, Domestic Production, Index 1982=100	Durable Goods, Thousands of Persons	Nondurable goods, Thousands of Persons	Construction, Thousands of Persons	Financial Activities, Thousands of Persons	Government, Thousands of Persons	Retail Trade, Thousands of Persons	Wholesale Trade, Thousands of Persons	Other Services, Thousands of Persons	Professional and Business Services, Thousands of Persons	Leisure and Hospitality, Thousands of Persons	Information Services, Thousands of Persons	Education and Health Services, Thousands of Persons
	Code (FRED)	UNEMPLOY	PAYEMS	PPIACO	RNUSBIS	WPU0561	DMANEMP	NDMANEMP	USCONS	USFIRE	USGOVT	USTRADE	USWTRADE	USSERV	USPBS	USLAH	USINFO	USEHS
Table1: Data Definition	Variable	Unemployment	Total employment	Commodity price	Exchange rate	Oil price	Manufacturing	Manufacturing	Construction	Financial	Government	Trade	Trade	Services	Services	Services	Services	Services

Tables and Figures

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	Model 1	Model 2	Model 3	Model 4	Model 5	Model 6		
Share (%) of Sec. Shocks	CK: 2S	Full: 2S	CK: POil	Full: POil	$UN \setminus ER$	UN \ CP	CK: 7S	Full: 7S
Short-run Restrictions	5.1	3.9	18.1	32.6	63.7	36.1	20.3	18.8
Long-run Restrictions	49.5	48.1	NA	NA	NA	NA	84.7	65.5
CK's short-run estimates	9	NA	43	NA	NA	NA	27	NA
CK's long-run estimates	51	NA	NA	NA	NA	NA	82	NA

Table 2: Variance Decomposition

CK refers to the sample period 1955M02 to 1994M12.

Full sample refers to the period 1955M02 to 2016M08.

2S (2 sectors) implies that the model contains manufacturing and non-manufacturing employment only.

7S implies that the model is augmented with 5 more sectors (construction, financial, government, trade and services).

POil implies that the model is augmented with oil prices.

UN implies that total employment has been replaced with the unemployment rate.

ER implies that the model is augmented with the exchange rate.

CP implies that the model is augmented with producer price series.

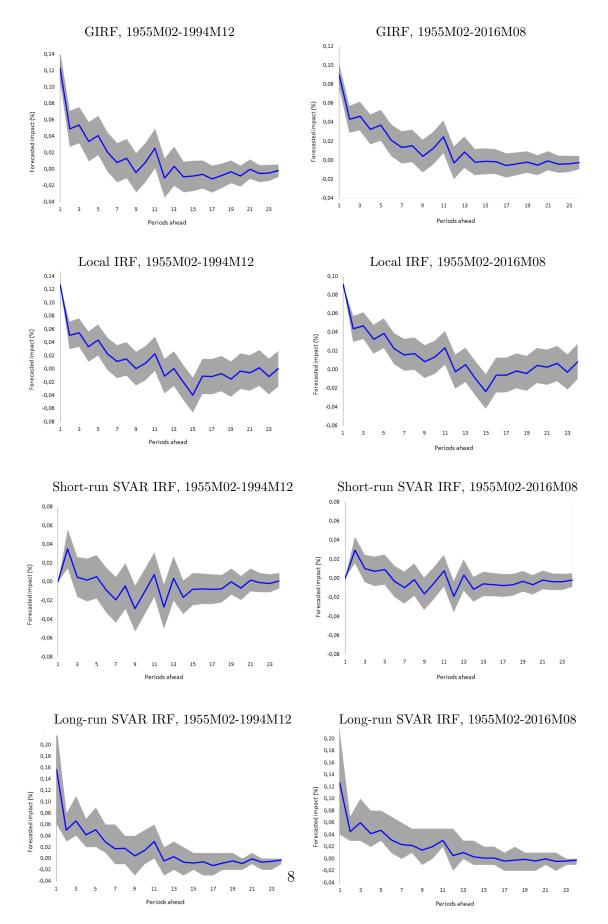


Figure 1: The response of total employment to reallocation shocks, two-sector specification

*The shaded areas for the GIRFs and the structural IRFs represent the ± 2 asymptotic S.E. confidence bands, while those for the local IRFs represent the 95% marginal ones.

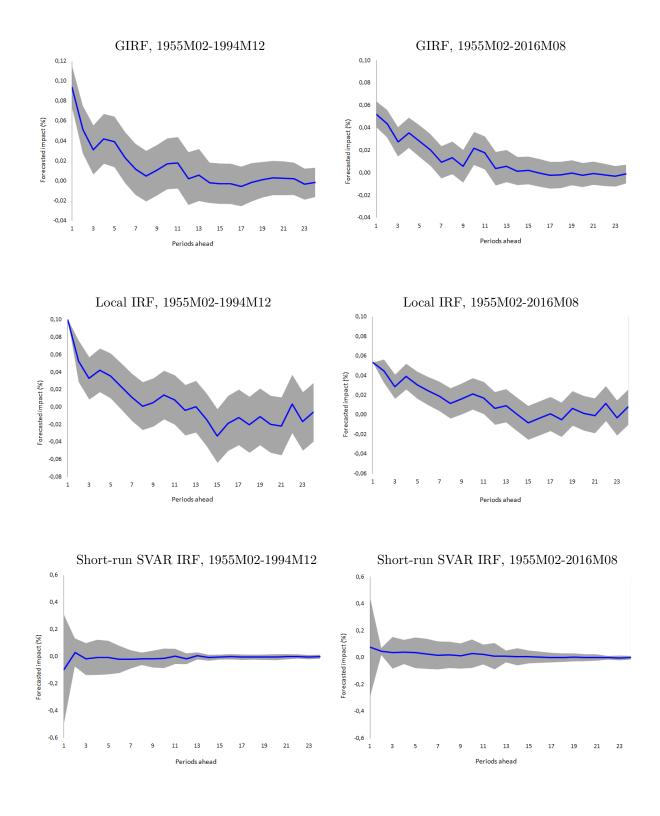
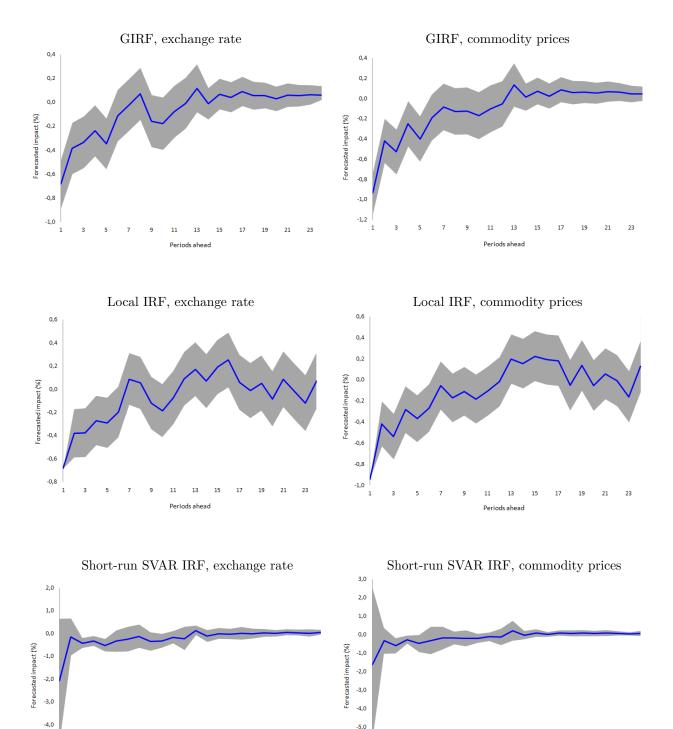


Figure 2: The response of total employment to reallocation shocks, oil prices augmented $$\rm V\!AR$$

Figure 3: The response of total unemployment to reallocation shocks, exchange rate / commodity prices augmented VAR



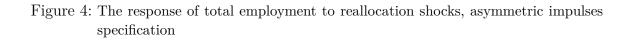


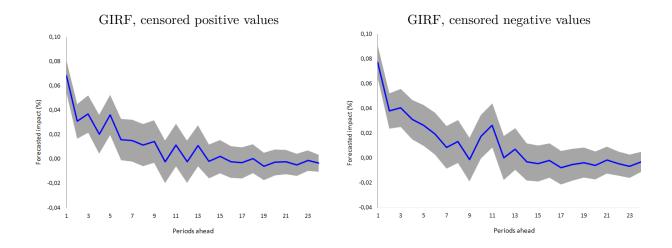
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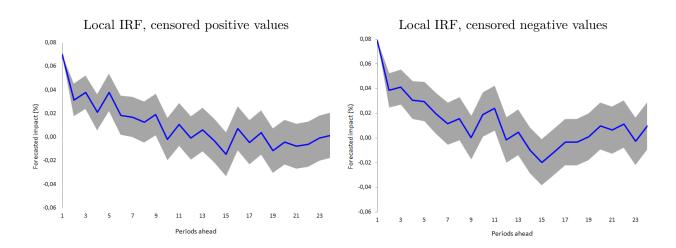
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11 13 15 17 19

Periods ahead







Appendix

1. Short-run restrictions

We impose short-run restrictions by following Sims's (1980) recursive identification scheme, so that the structural shocks u_{it} are just identified and recovered from the reduced ε_{it} . As in the AB model of Amisano and Giannini (1997) we have:

$$\begin{bmatrix} 1 & \alpha_{12} \\ \alpha_{21} & 1 \end{bmatrix} \begin{bmatrix} \varepsilon_{1\tau} \\ \varepsilon_{2\tau} \end{bmatrix} = \begin{bmatrix} b_{11} & 1 \\ 1 & b_{22} \end{bmatrix} \begin{bmatrix} u_{1t} \\ u_{2t} \end{bmatrix} \implies \begin{array}{c} \varepsilon_{1t} = b_{11}u_{1t} \\ \varepsilon_{2t} = -a_{21}u_{1t} + b_{22}u_{2t} \end{bmatrix}$$

Under this scheme, sectoral shocks have no effects on aggregate employment in the current period.

2. Long-run restrictions

In line with Blanchard and Quah (1989), we impose long-run restrictions via the accumulated responses to structural innovations. Thus $C = \Psi_{\infty} A^{-1} B$, is the long-run response matrix, $\Psi = (I - A_1 - A_2 - \dots - A_{\rho})^{-1}$ are the estimated accumulated responses from reduced-form shocks, A_{ρ} is the coefficient matrix for the lagged values of the endogenous and B is the matrix of the simultaneous coefficients. Imposing $C_{2,1} = 0$ implies that total employment shocks have no long-run impact on the manufacturing sector.

3. Augmented SVAR

Consistently with CK, we expand the bivariate VAR with the inclusion of the log-difference of the price of crude petroleum (DZ). DZ would behave as a univariate autoregression and would not affect DN directly. These restrictions would identify the system (Hamilton, 1994, p 330-331). We depart from CK by augmenting the underlying bivariate VAR with a third equation describing the dynamic behavior of a commodity price index instead of oil prices.

4. Alternative Impulse Response Functions

We employ alternative impulse response functions (IRFs) that were not available at the time of the CK publication. These are derived by the coefficient matrices of the moving average representation of a VAR (IRF to orthogonalized shocks). We look at the generalized impulse responses (GIRF) of Pesaran and Shin (1998):

$$GI_{\chi}(n,\delta,\Omega_{t-1}) = E(\chi_{t+n}|\varepsilon_t = \delta,\Omega_{t-1}) - E(\chi_{t+n}|\Omega_{t-1})$$

where χ is the vector of endogenous, Ω_{t-1} is the information set, δ is the shock size and n is the horizon.

Furthermore, following Jorda (2005, 2009) we use local projections instead of extrapolations. This approach is more robust to misspecification and can accommodate nonlinear and flexible specifications. The forecasts of this dynamic analysis are estimated at each period of interest rather than using the same set of information for longer horizons. The impulse responses from the local projections can be written as:

$$IR(t, s, d_i) = B_1^s d_i$$

where d_i is the shock to the i^{th} variable of the VAR, B_1^s are matrices of coefficients for each lag of variable 1 and s is the horizon of the forecast.

5. Data

The dataset is extracted from the Federal Reserve Bank of St. Louis (originally from the Bureau of Labor Statistics). These are monthly observations for the employment shares of manufacturing, services, construction, financial activities, trade and government, total employment, price of crude petroleum, real effective exchange rate, producer price index and total unemployment. All series are seasonally adjusted, and the sample ranges from 1955M02 to 2016M08 containing 739 observations. Table 1 presents the variables.

6. Additional Figures

Here we present the impact the real exchange rate and commodity price have directly on unemployment (through simple GIRFs and local $IRFs^5$):

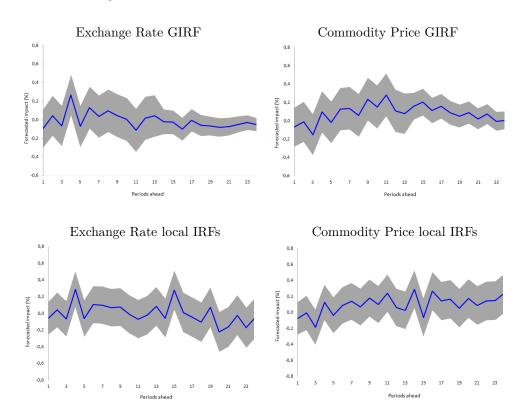


Figure 5: Responses of Unemployment to the indicated variables

⁵The structural case is equivalent with the simple GIRF.

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