# Research on the Relationship among the Total Factor Productivity of Three Industries in China

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**Abstract.** In order to study the relationship among the total factor productivity of three industries, the Vector Autoregression Model (VAR) was applyed in this paper on the basis of the existing literature research. The conclusion is that the secondary industry technology progress and the tertiary industry technology progress promote the primary industrial technology progress, the primary industry and the tertiary industry technological progress promote the technological progress of the secondary industry. The innovation of this paper is that it distinguished among three industries to study the relationship of the total factor productivity.

### Introduction

Many scholars have studied the total factor productivity from different perspectives (Machek Ondrej and Hnilica Jiri, 2012; Ben Soltane Bassem, 2014; Yen-Chun Chou and Benjamin B.M. Shao, 2014; Jungsoo Park, 2012; Almas Heshmati and Subal C. Kumbhakar, 2011). Alberto Colino et al.(2014) examined the determinants of total factor productivity growth in 26 OECD countries between 1965 and 2010. Sangeeta Pratap and CarlosUrrutia(2012) explored the role of financial frictions in exacerbating the misallocation of resources and explaining this drop in total factor productivity (TFP). Frank Asche et al.(2013) analyzed total factor productivity change in the Norwegian salmon aquaculture sector from 1996 to 2008. Yu Sheng and Ligang Song(2013) Used the firm-level census data to re-estimate the total factor productivity of firms in China's iron and steel industry and examine its potential determinants over the period 1998-2007. Kok Fong See and Fei Li(2015) examined the total factor productivity change of the UK airport industry from 2001 to 2009 using a Hicks-Moorsteen index method and got the conclusion that the industry experienced an average annual growth in TFP of 0.32 per cent.

Based on the existing literature research, this paper studied the relationship between the total factor productivity of the three industries in China from 1952 to 2013.

### Model, Index and Data

According to the Douglas production function, the relationship between the output and the input can

be expressed as:  

$$Y = AK^{\alpha}L^{\beta}\mu$$
(1)

Therefore, the calculation formula of total factor productivity can be obtained as follows:

$$TFP_t = \frac{Y_t}{K^{\alpha} L^{\beta}}$$
<sup>(2)</sup>

 $_{Y}$  refers to the total industrial output value, represented by actual GDP, which is deflated by GDP deflator.  $_{K}$  and  $_{L}$  refer to the input of capital and labor.  $_{\alpha}$  and  $_{\beta}$  refer to the output elasticity of capital and labor respectively.  $_{\mu}$  refers to random perturbation term. Taking logarithm of TFP, this paper get LNPTFP as the index of the total factor productivity of the primary industry, and get LNSTFP as the index of the total factor of the secondary industry, LNTTFP as the index of the total factor productivity of the tertiary industry. Data in this paper are derived from the CSMAR database, the website of the Nationgal Bureau of Statistics of the People's Republic of China, the *New China 60 Years Statistical Data Compilation* and the "Quantitative Calculation Method on the Role Of

Scientific and Technological Progress in Economic Growth" issued by the State Planning Commission of China in 1992.

#### **Empirical Analysis**

**Stationary Test.** In this paper, the ADF method is used to test stationarity of the total factor productivity of each industry. The test results are shown in table 1.

	Table1 variables Stationarity lest					
Test	Test	ADF test	The critica	The critical value of each significant		
sequence	form	statistic		level		
	(C,T,K)		1%	5%	10%	
LNPTFP	(C,T,0)	-2.159173	-4.115684	-3.485218	-3.170793	Unstatationary
DLNPTFP	(C,N,0)	-6.071282***	-3.544063	-2.910860	-2.593090	Stationary
LNSTFP	(C,T,1)	-2.550908	-4.118444	-3.486509	-3.171541	Unstatationary
DLNSTFP	(C,N,1)	-9.040902***	-3.546099	-2.911730	-2.593551	Stationary
LNTTFP	(C,T,0)	-1.269348	-4.115684	-3.485218	-3.170793	Unstatationary
DLNTTFP	(C,T,0)	-6.711633***	-4.118444	-3.486509	-3.171541	Stationary

**Lag Test.** According to the test results of Table 1, all the variables are 1 stage single integration, therefore can be tested by the cointegration test method. Through test, the variables are not cointegrated relationship. Therefore, the VAR model is constructed to analyze the relationship among the difference of these variables. The optimal lag period of the VAR model is determined in table 2.

Table2	Variables	Lag	Length	Test

			U	U		
Lag	LogL	LR	FPE	AIC	SC	HQ
0	186.0048	NA	3.65e-07	-6.310512	-6.203937*	-6.268999
1	202.6125	30.92466	2.81e-07	-6.572846	-6.146547	-6.406794
2	215.2656	22.25193*	2.48e-07*	-6.698813*	-5.952791	-6.4082223*
3	222.3824	11.77962	2.67e-07	-6.633877	-5.568131	-6.218748

Table3 Test Results of VAR Model					
	DLNPTFP	DLNSTFP	DLNTTFP		
DLNPTFP(-1)	0.262893	0.586038	0.218542		
	(0.13994)	(0.19808)	(0.13459)		
	[1.87859]	[2.95853]	[1.62372]		
DLNPTFP(-2)	-0.183736	-0.101307	0.067111		
	(0.13803)	(0.19538)	(0.13276)		
	[-1.33110]	[-0.51850]	[0.50551]		
DLNSTFP(-1)	0.184455	-0.367784	-0.129232		
	(0.09765)	(0.13822)	(0.09392)		
	[1.88894]	[-2.66084]	[-1.37600]		
DLNSTFP(-2)	-0.041195	-0.419444	-0.034880		
	(0.09675)	(0.13694)	(0.09305)		
	[-0.42581]	[-3.06292]	[-0.37486]		
DLNTTFP(-1)	0.208330	0.412329	0.435929		
	(0.15646)	(0.22147)	(0.15048)		
	[1.33152]	[1.86182]	[2.89691]		
DLNTTFP(-2)	0.211960	-0.294157	-0.088190		
	(0.15769)	(0.22321)	(0.15167)		
	[1.34411]	[-1.31783]	[-0.58146]		
С	0.028213	0.033895	-0.001478		
	(0.01167)	(0.01652)	(0.01123)		
	[2 41702]	[2 05148]	[-0.13146]		

F Statistics	3.101418	5.283126	2.583148
AIC		-6.748249	
SIC		-6.008787	

**VAR Model.** According to table 2, there are four criteria of which the optimal lag period is 2 stage in the five criteria of LR, FPE, AIC, SC and HQ. Therefore, the 2 stage lag is chosen to construct the VAR model. The test results are shown in table 3. The number in the parentheses is the standard error and the T statistics in the brackets.

**Granger Causality Tests.** The granger causality tests of DLNPTFP, DLNSTFP and DLNTTFP are shown in table 4, table 5 and table 6.

<u> </u>					
Lag	Null Hypothesis	F Statistics	Prob.	Test Result	
1	DLNSTFP does not Granger Cause DLNPTFP	7.63154	0.0077	Reject	
				č	
	DLNPTFP does not Granger Cause DLNSTFP	3.91381	0.0527	Reject	
	6			5	
2	DLNSTEP does not Granger Cause DLNPTEP	4.75934	0.0125	Reject	
-			0.0120	10,000	
	DLNPTEP does not Granger Cause DLNSTEP	4 67736	0.0134	Reject	
	DEI (1 111 does not Granger Gause DEI (5 111	1.07750	0.0151	Reject	
3	DLNSTEP does not Granger Cause DLNPTEP	3 13789	0.0332	Reject	
5		5.15705	0.0002	regeer	
	DI NPTEP does not Granger Cause DI NSTEP	4 08014	0.0113	Reject	
		7.00017	0.0115	Reject	

Table4 Granger Causality Tests of DLNPTFP and DLNSTFP

#### Table5 Granger Causality Tests of DLNPTFP and DLNTTFP

Lag	Null Hypothesis	F Statistics	Prob.	Test Result
1	DLNTTFP does not Granger Cause DLNPTFP	7.04475	0.0103	Reject
	DLNPTFP does not Granger Cause DLNTTFP	5.07970	0.0281	Reject
2	DLNTTFP does not Granger Cause DLNPTFP	5.34344	0.0076	Reject
	DLNPTFP does not Granger Cause DLNTTFP	2.73052	0.0742	Reject
3	DLNTTFP does not Granger Cause DLNPTFP	3.18296	0.0315	Reject
	DLNPTFP does not Granger Cause DLNTTFP	2.33244	0.0850	Reject

#### Table6 Granger Causality Tests of DLNSTFP and DLNTTFP

Lag	Null Hypothesis	F Statistics	Prob.	Test Result
1	DLNTTFP does not Granger Cause DLNSTFP	1.94684	0.1683	Accept
	DLNSTFP does not Granger Cause DLNTTFP	3.14823	0.0813	Reject
2	DLNTTFP does not Granger Cause DLNSTFP	2.13560	0.1280	Accept
	DLNSTFP does not Granger Cause DLNTTFP	1.69252	0.1937	Accept
3	DLNTTFP does not Granger Cause DLNSTFP	1.55192	0.2124	Accept
	DLNSTFP does not Granger Cause DLNTTFP	1.96056	0.1316	Accept

#### Summary

From the relationship between industrial production efficiency, the secondary industry technology progress and the tertiary industry technology progress promote the primary industrial technology progress. The primary industry and the tertiary industry technological progress promote the technological progress of the secondary industry. The primary industry technology progress promotes the technological progress of the tertiary industry. The secondary industry technological progress

progress and the technological progress of the tertiary industry have been changed in a negative direction. The primary industrial production efficiency has a significant effect on the secondary industry and the tertiary industry production efficiency. The effect of the secondary industry production efficiency is not more significant than the effect of the production efficiency of the primary industry, and the VAR model results show that the efficiency of the secondary industry is likely to have a reverse effect on the production efficiency of the tertiary industry, which possibly because that the resource allocation in the secondary industry production efficiency of the tertiary industry.

Therefore, the improvement of the production efficiency of the primary industry is the basis of the improvement of the secondary industry production efficiency and the tertiary industry production efficiency to improve the production efficiency of the secondary and the tertiary industry. The secondary industrial production efficiency in the short term improve the production efficiency of the primary industrial mechanization and modernization can not do without the support of the secondary industry. Therefore, we should strive to play the role of the secondary industry development improving the efficiency of the primary industrial production efficiency to improve the degree of agricultural modernization, so as to improve the efficiency of agricultural production. Because it is possible that the improvement of the secondary industry, so it should be more focused on the support of the tertiary industries.

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