Impact of lack of sufficient incentives on the healthcare provision of township healthcare centers in rural China

Zhongliang Zhoua,§, Chi Shena,§, *, Sha Laiab, Wanyue Dongb,
Yaxin Zhaoa, Dan Caoa, Dantong Zhaoa, Yangling Renab, Xiaojing Fanab

aSchool of Public Policy and Administration, Xi’an Jiaotong University, Xi’an, China, 710049
bSchool of Health Economics and Management, Nanjing University of Chinese Medicine, Nanjing, China, 210023
cSchool of Public Health, Health Science Center, Xi’an Jiaotong University, Xi’an, China, 710061

§ Zhongliang Zhou and Chi Shen contributed equally to this work.

*Corresponding author: Chi Shen, PhD, School of Public Policy and Administration, Xi’an Jiaotong University, No.28 Xianning West Road, Xi’an, 710049, China. Email: shenchi@outlook.com.
Zhongliang Zhou, zzliang1981@163.com
Chi Shen, shenchi@outlook.com
Sha Lai, shayanglala0217@163.com
Wangyue Dong, wanyuedong@foxmail.com
Yaxin Zhao, zyaxin1996@163.com
Dan Cao, cdan1996@163.com
Dantong Zhao, zhaodt1224@163.com
Yangling Ren, ryling0317@163.com
Xiaojing Fan, emirada@163.com
Abstract

Background
China's government launched a large-scale healthcare reform from 2009. One of the main targets of this round reform was to improve the primary health care system. Major reforms for primary healthcare institutions include increasing government investment. However, it is lack of an empirical study based on large sample to catch long-term effect of increased government subsidy and lack of sufficient incentives on township healthcare centers (THCs), therefore, this study aims to fill this gap by conducting an empirical analysis of THCs in Shaanxi province in China.

Methods
We collected nine years (2009 to 2017) data of THCs from the Health Finance Annual Report System (HFARS) that was acquired from the Health Commission of Shaanxi Province. We applied two-way fixed effect model and continue difference-in-difference (DID) model to estimate the effect of percentage of government subsidy on medical provision.

Results
A clear jump of the average percentage of government subsidy to total revenue of THCs can be found in Shaanxi province in 2011, and the average percentage has been more than 60% after 2011. Continue DID models indicate every 1% percentage of government subsidy to total revenue increase after 2011 resulted in a decrease of 1.1% to 3.5% in THCs healthcare provision (1.9% in medical revenue, 1.2% in outpatient visit, 3.5% in total occupy beds of inpatient, 1.1% in surgery revenue, 2.1% in sickbed utilization rate). The results show that the THCs with high government subsidy reduce the number of medical services after 2011.

Conclusions
We think that it is no doubt that the government should take more responsibility for the financing
of primary healthcare institutions, the problem is when government plays a central role in the financing and delivery of primary health care services, more effective incentives should be developed.

**Keyword**: Incentives, Township healthcare centers, Healthcare reform, China

**Background**

In order to deal with the problem of the healthcare system and achieve the goal of establishing a universal coverage healthcare system for all citizens by 2020, the Chinese government launched a large-scale healthcare reform in 2009 [1, 2]. Policies in this round reform can be summarized into five parts: Social health security, Essential medicines, Primary healthcare, Essential public health services program, Public hospitals [3, 4]. To improve the performance of the primary healthcare system is one of the main targets of this round reform. Major reforms for primary healthcare institutions (PHIs) include increasing government investment, eliminating drug mark-ups, and general practitioners contract service [4–6].

Increasing government investment means the government provides financial subsidies to cover all the costs of PHIs, such as construction and equipment expenditure, personnel salary, and operational expenses for public health services [7]. Before the reform, the PHIs' cost was financed by the government and revenue of PHIs (including drug sales) [8]. According to increasing government investment and eliminating 15% drug mark-ups, China government hopes the PHIs can provide convenient and low-cost essential health services for residents rather than profit-oriented [1].

Along with the change of compensation strategy, the incentives of primary healthcare staff are also changed. After the reform, the total salary of staff in township healthcare centers (THCs) is
approved based on the number of staff and service workload, referring to the average salary of the local government staff [7]. The salary consists of basic salary and performance-based bonus. The basic salary shares 60% to 70% of the total salary, performance-based bonus accounts for 30% to 40% of total salary [8]. However, every staff can get 70% of the bonus, only 30% of the bonus is really based on performance [9]. Therefore, the low bonus based on performance leads to a problem - a lack of sufficient incentives.

Due to professional ethics, practice norms, and financial incentives affect the performance of staff in THCs, a lack of sufficient incentives would lead to inefficiencies and poor quality of healthcare services [10]. From 2009 to 2017, the percentage of outpatient services provided by THCs decreased from 16.0% to 13.6%, and inpatients services decreased from 28.7% to 16.6% [11]. One study shows that the compensation and incentive strategy is the key point in the reform of THCs [12]. Another study conducted in Anhui province in China shows that healthcare services provided by THCs with high government subsidy and performance-based salary system have reduced to varying degrees between 2009 and 2010 [9]. In Hubei province in China, the technical efficiency of THCs decreased from 2012, and THCs experienced a decline in productivity [13]. However, existed research have some limitations: small sample (6 and 48 THCs) [9, 13], descriptive analysis, or theoretical analysis [9, 12].

It is lack of an empirical study based on a large sample to catch the long-term effect of increased government subsidies and lack of sufficient incentives on THCs. This study aims to estimate the effect of the percentage of government subsidy on total revenue on the medical service of THCs. We conducted an empirical analysis by using a balanced panel data that includes 1199 THCs from 2009 to 2017 in Shaanxi province, China. We find that higher percentage of government subsidy to total revenue of THCs, less provision of medical services, lack of sufficient
incentives made THCs shrink in terms of medical service.

Box 1 The entire history of China’s township healthcare centers

The main component of PHIs in rural China is township health centers (THCs). The entire history of China's THCs is about how to balance the relationship between government and market in terms of financing (Figure.1). At the stage of the People's Republic of China funded (1958 - 1980), the government strongly supported and financed the THCs, and people's health statue intensely promoted during this period [14]. At the stage of Economic Reform and Open Up (1981 - 2001), the government changed the financing policy and made the THCs into the market, which caused a massive loss of technical staff from THCs and resident’s healthcare utilization in THCs decreased [14–16]. After aware of the shrinking of THCs, the government began to increase the subsidy to THCs from 2002 and continue to increase it from 2009 [7].

   [Insert Figure 1 about here]

Methods

Study site and data collection

Shaanxi province is located in northwest China and has an area of 205,800 square kilometers, the total population in 2017 is over 38.35 million [17]. There are 11 prefecture-level cities, 77 counties [18]. The economic and development level of Shaanxi province is the average level of China.

We collected nine years (2009 to 2017) data of THCs from the Health Finance Annual Report System (HFARS) that was acquired from the Health Commission of Shaanxi Province. The HFARS includes resource, service, and finance information of each health institutions hosted by the government. The data include number of medical personnel, number of beds, number of visits
in hospitals, and day of in-hospital patient bed occupancy, revenue, medical revenue, and more other provision indicators. Due to changes in administrative divisions and the elimination of missing values, finally, this study included 1199 THCs from total 1552 THCs and constructed a balanced panel dataset [19].

**Indicators and models**

We chose the percentage of government subsidy to total revenue of THCs as the explanatory indicator that can identify the degree of compensation of THCs from the government. Considering the lack of efficient incentives would firstly affect the medical services provided by THCs, we selected the indicators that can reflect the amount of inpatient and outpatient services as the explained variable. Moreover, some confounders will affect the THCs’ provision, such as number of physicians and sickbeds, we chose these confounders as control variables. The detail about variables including in this study can be found in Table.1.

[Insert Table.1 about here]

A two-way fixed-effect model was used to estimate the average treatment effect of the percentage of government subsidy, estimating equation (1) is written as:

\[ y_{it} = \beta \cdot Subsidy_{it} + z_i \delta + \lambda t + \mu_i + \epsilon_{it} \]  

(1)

Where,

- where \( i \) indexes THCs and \( t \) indexes years.
- \( y_{it} \) is the indicator that can reflect the amount of inpatient and outpatient services of THCs.
- \( Subsidy_{it} \) is the government subsidy as % of total revenue of \( i \) THCs at \( t \) time.
- \( z_i \) is the control variable.
- \( \lambda t \) is the year fixed effect.
- \( \mu_i \) is the THCs fixed effect.
• $\varepsilon_{it}$ is the residual error.

According to the official document [20], released on July 6, 2011, government subsidy of THCs increased from 2011 in Shaanxi province. Since there is a clear policy cut-point, we also applied a difference-in-difference (DID) model to compare the average effect of subsidy percentage on the provision of medical services of THCs before and after 2011. However, the DID model used in this study is different from standard DID model [21], the difference is that the intensity of treatment is a continuous measure (i.e., government subsidy as % of total revenue). The detail about the continuous DID model can be found here [22]. Estimating equation (2) is written as:

$$y_{it} = \beta \cdot Subsidy_{it} \cdot Year^{2011} + z_t \delta + \lambda t + \mu_t + \varepsilon_{it}$$  \hspace{1cm} (2)$$

Where,

• where $i$ indexes THCs and $t$ indexes years.
• $y_{it}$ is the indicator that can reflect the amount of inpatient and outpatient services of THCs.
• $Subsidy_{it}$ is the government subsidy as % of total revenue of $i$ THCs at $t$ time.
• $z_t$ is the control variable.
• $\lambda t$ is the year fixed effect.
• $\mu_t$ is the THCs fixed effect.
• $\varepsilon_{it}$ is the residual error.
• $Year^{2011}$ is a dummy variable (2009 and 2010 = 0, after 2011 = 1)

As we found, the distribution of explained variables is over discrete in the process of descriptive analysis, and we made a logarithmic conversion of explained variables in models. All analysis in this study was performed by R 3.5.3 [23].
Results

Basic information of explained and explanatory variables

We can find a clear jump of the average percentage of government subsidy to total revenue of THCs in Shaanxi province in 2011 from Figure.2, and the average percentage has been more than 60% after 2011. In terms of medical services, medical revenue, and inpatient and outpatient services increased from 2010 to 2017, and there was a decrease in surgical income and no increase in sickbed utilization rate (Table.2). Furthermore, Figure.3 presents the scatter plots of THCs between government subsidy to total revenue and total occupy beds of inpatient yearly from 2009 to 2017, the locally estimated scatterplot smoothing (LOESS) curves show that the relationship between government subsidy and total occupy beds of inpatient are negative. Other indicators have the same relationship, please see Figure.A1 to Figure.A4 in Additional file).

Two-way fixed effect model result

Based on the results from scatter plots, we used two-way fixed effect models to estimate the average effect of the percentage of government subsidy to total revenue on number of medical services. We performed a Hausman Test to compare fixed effect model with random effect model, the result indicates that fixed effect model is better. Table.3 shows that all the coefficients of the percentage of government subsidy are negative and significant in model 1 to model 5, which indicates that every 1% percentage of government subsidy to total revenue increase resulted in a decrease of 1.3% to 2.6% in THCs healthcare provision (2.4% in medical revenue, 1.3% in outpatient visit, 2.6% in total occupy beds of inpatient, 1.4% in surgery revenue, 1.8% in sickbed utilization rate).

[Insert Figure.2, 3 and Table.2 about here]

[Insert Table.3 about here]
Continue difference-in-difference model result

In order to evaluate the average effect of the percentage of government subsidy to total revenue on number of medical services before and after 2011, we applied continue DID model. Table.4 shows that coefficients of interaction between the year dummy variable and the percentage of government subsidy to total revenue are also negative, the coefficients indicate every 1% percentage of government subsidy to total revenue increase after 2011 resulted in a decrease of 1.1% to 3.5% in THCs healthcare provision (1.9% in medical revenue, 1.2% in outpatient visit, 3.5% in total occupy beds of inpatient, 1.1% in surgery revenue, 2.1% in sickbed utilization rate). The results show that the THCs with high government subsidy reduce the number of medical services after 2011.

[Insert Table.4 about here]

Discussion

Our study indicates that, from 2009 to 2017, medical services provided by THCs in Shaanxi province slightly increased, and the government subsidy to THCs experienced substantial growth from 2011. However, the increase of government subsidy did not make the THCs provide more necessary medical services after completing the tasks set by the government. THCs did not meet the health needs of rural residents stimulated by New Rural Cooperative Medical [24, 25].

The findings in our study are similar to the previous studies conducted in Anhui and Hubei province [9, 13], the new contribution of our research is that we confirm the effect of compensation policy changes on the medical services provided by THCs from the perspective of large samples and long-term effects. So, is the incentive method important for primary healthcare institutions? Several review papers indicate that the financing policies for primary health care in China in this round reform might lead to low productivity in primary health-care institutions, and incentive
policy is one of the key elements in building an integrated primary health-care system for China [5, 26]. However, it is difficult to make an appropriate incentive strategy. An insurance intervention study conducted in Ningxia province in China revealed that changes in payment method on supply side failed to change residents’ healthcare seeking behavior [27].

The original intention of increasing government subsidy for THCs is to compensate for the shortage of income and expenditure of THCs when eliminating 15% drug mark-ups [7]. Along with the government subsidy increase, the incentive strategy shifts from profit-driven (e.g., prescribing diagnostic tests and drugs) [10] to “approved task, approved revenue and expenditure, performance-based bonus”, which means each THCs will be set an annual task. How to set task goals plays a key role in maintaining the quality and efficiency of health services in THCs. However, factually, the task is simply approved by reference to service population and average health service provision of the THCs over the past three years [20]. Under the situation that high government subsidy, unscientific task setting method, and lack of incentive in salary system, it is common sense that staffs in THCs have no incentives to do extra work after completing the task. Therefore, THCs firstly reduced high-risk medical services, such as hospitalization and surgery, we have reason to believe that THCs with higher government subsidy will reduce medical services compared to THCs with lower government subsidies.

There is a long-standing debate in China that the financing and provision of healthcare services should depend on the market or government. In the past ten years of reform, the Chinese government has invested a large number of funds in the primary healthcare system, but it has not achieved the expected outcomes. It is no doubt that the government should take more responsibility for the financing of primary healthcare institutions. The problem is that when the government plays a central role in the financing and delivery of primary health care services, more effective
incentives should be developed. Supply-side incentives can promote to build an integrated healthcare delivery system based on primary healthcare system in China [28]. Fortunately, however, the Chinese government has begun to promote the reform of the salary system of public hospitals, which proposes “two permits”, namely, allowing health institutions to break the limitation of the current level of salary and allowing health institutions to use the revenue balance as personnel reward [29].

There are two limitations in this study. Firstly, as it is difficult to collect the population of towns, we have to use the number of authorized beds of THCs as a proxy variable to control the influence of the population. Secondly, as we mentioned before, services provided by THCs include public health and medical service, however, only indicators of medical service were included in our study because of no suitable indicators about the provision of public health in HFARS.

**Conclusion**

Because of the changes in the compensation policy of township health centers and the lack of incentives, the increase of government subsidy did not make the township health centers to provide more necessary medical services, which indirectly leads to a decline in medical service quality and aggravates the problem of “KAN BING NAN”. It is no doubt that the government should take more responsibility for the financing of primary healthcare institutions. The problem is that when the government plays a central role in the financing and delivery of primary health care services, more effective incentives should be developed.

**Abbreviations**

DID: Difference-in-Difference;

HFARS: Health Finance Annual Report System;
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Authors' contributions

ZL Z and CS proposed the study framework and analysis strategy. ZL Z wrote first draft of the manuscript. YL R, YX Z and WY D carried out all aspects of data collection and data analysis. DC and DT Z provided the analysis assistance. SL and XJ F gave many suggestions about the study design. CS and ZL Z gave the revised suggestions and approved the final version.

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Availability of data and materials
The datasets used in this study are available from the corresponding author on reasonable request.

**Declarations**

**Ethics approval and consent to participate**
Health Finance Annual Report System (HFARS) only includes statistical information of healthcare institutions, there is no human ethical parameter in our data of THCs.

**Consent for publication**
Not applicable.

**Competing interests**
Zhongliang Zhou is a member of the editorial board of BMC Health Services Research journal.
The authors declare no other conflicts of interest.

**Author details**
aSchool of Public Policy and Administration, Xi’an Jiaotong University, Xi’an, China, 710049
bSchool of Health Economics and Management, Nanjing University of Chinese Medicine, Nanjing, China, 210023
cSchool of Public Health, Health Science Center, Xi’an Jiaotong University, Xi’an, China, 710061

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Table 1. Detail about indicators in this study

| Dependent variables                  | Control variables                  | Explanatory variable                                      |
|--------------------------------------|------------------------------------|-----------------------------------------------------------|
| Medical revenue                      | Number of physician                | Government subsidy as % of total revenue                  |
| Total occupy beds of inpatient       | Type of THCs (Center or General)   |                                                           |
| Total outpatient visits              | Number of authorized bed           |                                                           |
| Surgery revenue                      | Separation between revenue and expenditure |                                                      |
| Sickbed utilization rate             |                                    |                                                           |

Table 2. Summary of dependent variables

| Year | Observation | Medical revenue (RMB) | Outpatient visit | Total occupy beds of inpatient | Surgery revenue (RMB) | Sickbed utilization rate (%) |
|------|-------------|-----------------------|------------------|-------------------------------|-----------------------|-----------------------------|
|      |             | Mean                  | SD               | Mean                          | SD                    | Mean                         | SD                        |
| 2009 | 1199        | 234326.1              | 317225.4         | 11735.5                       | 11280.2               | 2413.9                       | 2673.1                   |
| 2010 | 1199        | 260552.9              | 354245.0         | 11959.4                       | 11280.2               | 2381.2                       | 2759.5                   |
| 2011 | 1199        | 746156.5              | 790213.3         | 12356.4                       | 12128.8               | 2549.2                       | 2984.1                   |
| 2012 | 1199        | 97187.6               | 100663.1         | 13097.8                       | 12697.1               | 3253.2                       | 3777.8                   |
| 2013 | 1199        | 1088195.9             | 1073457.4        | 13260.8                       | 12647.3               | 3356.3                       | 3767.9                   |
| 2014 | 1199        | 1172641.7             | 1141956.2        | 13598.3                       | 12306.6               | 3277.1                       | 3637.1                   |
| 2015 | 1199        | 1258281.4             | 1218177.7        | 14016.6                       | 13175.1               | 3342.5                       | 3894.4                   |
| 2016 | 1199        | 1356923.4             | 1312443.4        | 14023.1                       | 12760.6               | 3408.1                       | 4196.3                   |
| 2017 | 1199        | 1617121.6             | 1601525.4        | 14498.5                       | 12989.5               | 3886.7                       | 4615.6                   |
| Subsidy percentage       | -0.024*** | -0.013*** | -0.026*** | -0.014*** | -0.018*** |
|-------------------------|-----------|-----------|-----------|-----------|-----------|
|                         | (<0.001)  | (<0.001)  | (0.001)   | (0.003)   | (0.001)   |
| Type of THCs [C220]     | -0.022    | -0.002    | 0.074     | 0.071     | -0.038    |
|                         | (0.036)   | (0.038)   | (0.109)   | (0.226)   | (0.068)   |
| Number of Physician     | 0.005***  | 0.005***  | 0.005**   | 0.011**   | 0.002     |
|                         | (0.001)   | (0.001)   | (0.002)   | (0.005)   | (0.002)   |
| Number of authorized bed| 0.006***  | 0.005***  | 0.034***  | 0.018***  | -0.000    |
|                         | (0.001)   | (0.001)   | (0.003)   | (0.006)   | (0.002)   |
| Separation revenue policy| -0.054***| -0.085***| -0.228***| -0.290** | -0.108*** |
|                         | (0.020)   | (0.022)   | (0.061)   | (0.128)   | (0.039)   |
| Individual effect       | YES       | YES       | YES       | YES       | YES       |
| Time effect             | YES       | YES       | YES       | YES       | YES       |

Observations 10,791 10,791 10,791 10,791 10,791

$R^2$ 0.264 0.097 0.063 0.006 0.052

Adjusted $R^2$ 0.171 -0.018 -0.055 -0.120 -0.068

$F$ Statistic (df = 5; 9579) 688.720*** 204.784*** 129.574*** 10.739*** 105.410***

Note: ***p<0.01; **p<0.05; * p<0.1

p value of Hausman Test is < 0.05, which indicates that fixed effect model is better
Table 4 Result of continuous measure difference-in-difference model

|                         | log(medical revenue) | log(outpatient visit) | log(total occupy beds of inpatient) | log(surgery revenue) | log(sickbed utilization rate) |
|-------------------------|----------------------|-----------------------|------------------------------------|----------------------|------------------------------|
| **Subsidy percentage*Year** | (1) -0.019***       | (2) -0.012***        | (3) -0.035***                      | (4) -0.011***        | (5) -0.021***                |
|                         | (<0.001)             | (<0.001)              | (0.001)                            | (0.003)              | (0.001)                     |
| Type of THCs [C220]     | 0.020                | 0.018                 | 0.093                              | 0.096                | -0.020                      |
|                         | (0.039)              | (0.039)               | (0.107)                            | (0.227)              | (0.068)                     |
| Number of Physician     | 0.005***             | 0.005***              | 0.005*                            | 0.011***             | 0.002                       |
|                         | (0.001)              | (0.001)               | (0.002)                            | (0.005)              | (0.002)                     |
| Number of authorized bed| 0.005***             | 0.004***              | 0.029***                           | 0.017***             | -0.003*                     |
|                         | (0.001)              | (0.001)               | (0.003)                            | (0.006)              | (0.002)                     |
| Separation revenue policy| -0.066***            | -0.090***             | -0.227***                          | -0.297***            | -0.110***                   |
|                         | (0.022)              | (0.022)               | (0.061)                            | (0.128)              | (0.038)                     |
| Individual effect       | YES                  | YES                   | YES                                | YES                  | YES                         |
| Time effect             | YES                  | YES                   | YES                                | YES                  | YES                         |
| Observations            | 10,791               | 10,791                | 10,791                             | 10,791               | 10,791                      |
| R²                      | 0.151                | 0.075                 | 0.088                              | 0.004                | 0.066                       |
| Adjusted R²             | 0.044                | -0.042                | -0.027                             | -0.122               | -0.052                      |
| F Statistic (df = 5; 9579)| 341.503***          | 154.581***            | 184.786***                         | 8.180***             | 135.773***                  |

Note: ***p<0.01; **p<0.05; * p<0.1

p value of Hausman Test is < 0.05, which indicates that fixed model is better

**Figure Legend**

Figure 1: Stages of development of township health centers
Figure 2 Tread of percentage of subsidy of THCs from 2009 to 2017

Figure 3 Relationship between subsidy and in-hospital from 2009 to 2017
Figure A1: Relationship between subsidy and medical revenue from 2009 to 2017
Figure A2 Relationship between subsidy and outpatient visit from 2009 to 2017
Figure A3: Relationship between subsidy and surgery revenue from 2009 to 2017
Figure A4 Relationship between subsidy and sickbed utilization rate from 2009 to 2017