Measuring energy security performance within China: Toward an interprovincial prospective

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Abstract:

China has been the world's largest energy consumer and producer for many years, yet while

myriad studies have investigated Chinese performance on energy metrics compared to other

countries, few to none have looked internally at Chinese provinces. This paper firstly develops a

five-dimensional evaluation system centered on the energy security dimensions of availability and

diversity, affordability and equality, technology and efficiency, environmental sustainability, and

governance and innovation. It then correlates these dimensions to 20 distinct energy security

metrics that are used to assess the energy security performance of 30 Chinese provinces, divided

into eight regions. Our results reveal both trends in energy policy and practice as well as provincial

status of comparative energy security for the year 2013. We find, for instance, that there is no

province which performs well in all five of the energy security dimensions, and that all provinces

confronted threats related to energy availability and diversity. We also demonstrate that in

comparative terms, the Middle Reaches of Yellow River and the Northwest were the most energy-

secure, while the Middle Reaches of Yangtze River and the Northeast were least energy-secure.

Keywords: energy security; Fuzzy AHP; PROMETHEE; sensitivity analysis; China

1. Introduction

Recently, there has been no shortage of scholarly or practitioner interest in energy security. Of particular concern is China, which surpassed the United States as the world's largest energy consumer in 2010, with a share of 20.3 percent of world total energy consumption that year (British Petroleum, 2011) [1]. Energy demand in China is expected to increase 60% further from 2015 to 2035, and by the early 2030s, China could become the world's largest energy importer, overtaking Europe in terms of volume of imports, with its import dependence rising from 15% to 23% (British Petroleum, 2015) [2]. China even became a net coal importer since 2009 (Rong and Victor, 2011) [3], and the country's dependence on foreign oil exceeded 60 percent in 2014 (Ministry of Land and Resources of PRC, 2014) [4]. These resource dependence concerns also do little to address a second serious problem of climate change, which requires less fossil fuel consumption for climate change mitigation (Jacobson, 2002) [5]. What's more, the consumption of fossil fuels causes serious ambient air pollution, leading to urban haze in major cities in eastern China, which has greatly threatened public health (Chen, et al., 2013) [6]. Further complicating matters, discussions about peak oil, price fluctuations, and energy inequality have drawn the attention of policymakers and investors, as energy security is closely related with national goals of sustainable development and economic growth (Sovacool and Brown, 2010) [7].

Indeed, as the third largest entity of the world, China is also the largest energy producer, with a share of 19% of world's energy production (U.S. Energy Information Administration, 2015) [8]. However, China's provinces vary greatly in energy endowment, economy development, industrial structure, technology development, and even social and cultural customs, and there is a great spatial disparity between flows of energy fuels and services within them (Chen et al, 2010; Wang et al., 2011; Bao and Fang, 2013; Zhang et al., 2012; National Bureau of Statistics of China, 2013)

- [9-13]. Due to these disparities, energy researchers and policymakers in China are left with at least three puzzling research questions:
 - (1) How can the spatial energy security trends of Chinese provinces be analyzed?
 - (2) How can energy security performance be measured at the provincial level?
 - (3) How can the provincial energy security of China be enhanced?

Interestingly, despite an abundance of research on energy security metrics and performance (Le Coq and Paltseva, 2009; Vivoda, 2010; Cohen et al., 2011; Sovacool et al., 2011; Sovacool, 2011, 2013a and 2013b) [14-20], no studies have as of yet answered these three questions. Admittedly, many studies have investigated China's energy security issues (Bambawale and Sovacool, 2011; Wu et al., 2012; Ren and Sovacool, 2014; Liu, 2014; Geng and Ji, 2014; Yao and Chang, 2014; Yang and Chen, 2014) [21-27]. For instance, Wu et al. (2012) measured China's security of energy supply with 14 indicators [22]. Geng and Ji (2014) [25] developed a multi-dimensional indicator system to evaluate China's national energy security. Yang and Chen (2014) [27] established an evaluation framework to measure Chinese performance on energy security metrics. None, however, have looked at Chinese energy security at the provincial scale.

To do so, this study first reviews the academic literature in energy studies to present five dimensions of energy security: availability and diversity, affordability and equality, technology and efficiency, environmental sustainability, and governance and innovation. It then corresponds these to 20 different metrics and assigns them weights based on Fuzzy Analytical Hierarchy Process (an established method in the energy studies and engineering fields), and investigates the energy security performance of 30 Chinese provinces with another two methods, Preference Ranking Organization Method for Enrichment Evaluations (PROMETHEE), and Sensitivity

Analysis. The final parts of the article present its results and analysis as well as policy implications and conclusions.

2. Dimensions and metrics of energy security

This section firstly reviews the conceptual underpinning of energy security, summarizing recent peer-reviewed research related to energy security dimensions and metrics.

2.1 Literature review

In the existing literature, energy security is most commonly defined as the reliable supply of energy at reasonable prices to support the economy and industry (Yergin, 1988; International Energy Agency, 2001; Bielecki, 2002; Dorian et al., 2006; Asif & Muneer, 2007; Jun et al., 2009) [28-33]. This traditional definition of energy security has been critiqued for being too narrow and for downplaying broader social and environmental factors such as climate change or community acceptance (Yergin, 2006) [34]. The "4A" framework of energy security (availability, accessibility, acceptability and affordability) proposed by APERC (Asia Pacific Energy Research Centre, 2007) [35] is a more representative conception, one utilized by numerous scholars (Hughes, 2012; Winzer, 2012) [36-37].

There is no shortage of simple or aggregated indexes to measure energy security. Kruyt et al. (2009) [38] overviewed 24 simple and aggregated indicators for security of supply found in literature. The most widely used are reserves to production ratios, diversity indices, including Herfindahl-Hirschman index (HHI) and Shannon Wiener index (SWI) (Neff. 1997; Vivoda, 2009; Cohen et al., 2011; Shakya and Shrestha, 2011) [16, 39-41]. A noteworthy energy security index developed by International Energy Agency (2007) [42] integrated HHI with the political stability

of importing sources and energy prices. Taking a similar approach, Lefèvre (2010) [43] designed an energy security price index and energy security physical availability index, and used these two indices to project the energy security of France and UK over the period of 2004–2030. Based on IEA's energy security index, Löschel et al. (2010) [44] developed ex-post and ex-ante indicators of energy security, and used them to depict the energy security of Germany, Netherlands, Spain and the United States. To measure the short-term risk of energy supply, Le Coq and Paltseva (2009) [14] developed a Risky External Energy Supply Index, which combined net import dependency, political risks of the supplying country, energy transport risks, energy fungibility and the economic importance of each energy type, and used it to measure the risks of oil, gas and coal for the EU members. By constructing the Gas Supply Security Index, Cabalu (2010) [45] combined gas intensity, net gas import dependency, ratio of domestic gas production to total domestic gas consumption and geopolitical risk, and examined the relative vulnerability to natural gas supply disruptions of seven gas-importing countries in Asia for year 2008. These indexes all make meaningful contributions in examining security of energy supply.

There are also a number of metrics of energy security that have been proposed in the literature from a more synthesized prospective. Vivoda (2010) [15] proposed a set of quantitative and qualitative indicators to evaluate national or regional energy security, supposing that it consists of 11 dimensions and 44 attributes. Von Hippel et al. (2011) [46] devised 29 indicators of energy security including six dimensions. Unfortunately, they failed to conduct any empirical study with their instrument for any country or region. Sovacool and Mukherjee (2011) [47] developed five dimensions consisting of 20 components and 320 simple indicators along with 52 complex indicators to assess energy security. Wu et al. (2012) measured China's energy supply security with 14 indicators [22]. Martchamadol and Kumar (2012) used 19 indicators, which can be

categorized into five sets, to analyze the energy security in Thailand for the period 1986-2030 in three energy scenarios presented by APERC [48].

2.2 Dimensions and metrics

To provide a comprehensive, yet capable and parsimonious approach to measuring energy on a provincial level, this paper synthesized such dimensions and indicators from the literature (Vivoda, 2010; Hippel et al., 2011; Sovacool and Mukherjee, 2011; Martchamadol and Kumar, 2012; Liu et al., 2012; Liu, 2014; Ren and Sovacool, 2014; Asdrubali, 2015) [15, 23-24, 46-50]. Thus, we hold that energy security consists of availability and diversity, affordability and equality, technology and efficiency, environmental sustainability, and governance and innovation. As presented in Table 1, these five dimensions can be effectively decomposed into 20 components, and thereby measured with 20 metrics. We must empathize that these five dimensions have been validated by recent quantitative assessments of energy security (Ren and Sovacool, 2014) [23] as well as qualitative research with energy experts (Sovacool et al., 2011) [17], a series of focus groups and workshops (Sovacool, 2013b) [20], and surveys of energy security attitudes both looking internally at China (Bambawale and Sovacool, 2011) [21] as well as a comparative sample of eleven countries including China (Sovacool et al., 2012; Sovacool, 2016) [51-52].

2.2.1 Availability and diversity

The dimension of availability and diversity represents security of energy supply and demand, which consists of factors relating to energy resources and endowments, energy production, imports and exports, and energy consumption. We decomposed this dimension into four components in Table 1: security of supply, energy potential, dependency, and diversity.

2.2.2 Affordability and equality

The dimension of affordability and equality refers to the economic aspects of energy security such as the prices of energy carriers, price fluctuations, and energy equality. Four components are included in this dimension in Table 1: stability, electricity generation costs, electricity equality, and affordability of gasoline.

2.2.3 Technology and efficiency

This dimension relates to the development of energy related technologies and overall energy efficiency. It comprises four components in Table 1: energy efficiency, grid efficiency, grid reliability, and capacity factors.

2.2.4 Environmental sustainability

The dimension of environmental sustainability represents the environmental and social aspects of energy production and consumption. We broke this dimension into five components in Table 1: land use, water pollution, climate change, acidification, and photochemical pollution.

2.2.5 Governance and innovation

The dimension of governance and innovation regards political factors related to energy security. We utilized three components in Table 1 to represent this dimension: market potential, innovation and research, and energy and environment management.

3. Data and methodology

With our dimensions and metrics chosen, we proceed to collect data on our metrics and then to weight them using the principles and procedures of Fuzzy AHP. After that, we use PROMETHEE to evaluate the energy security performance for 30 Chinese provinces.

3.1 Data collection and Processing

After establishing the metrics iterated above in Table 1, we proceeded to collect data on them for 30 provinces in China, excluding Hong Kong, Taiwan, Macao and Tibet. All data is for 2013, and is derived from the China Statistics Yearbook and the Statistics Yearbooks of the 30 provinces (2014) [53-83], excepting for the data on carbon emissions, which came from the reference (Bai, 2016) [84].

In order to achieve dimensionless analysis, data of the metrics was normalized by Eq. (1) and (2), respectively.

$$p_{ij} = (x_{ij} - \min x) / (\max x_j - \min x_j)$$
 (1)

$$p_{ij} = (\max x - x_{ij}) / (\max x_{i} - \min x_{j})$$
 (2)

where p_{ij} is the normalized value of the i_{th} province with respect to the j_{th} metric, x_{ij} represents the original value of the i_{th} province with respect to the j_{th} metric, and $\max x_{ij}$ and $\min x_{ij}$ are the maximum and minimum value of the 30 provinces with respect to the j_{th} metric.

3.2 Multi-criteria Decision Making Methods

To provide weights to the data and evaluate energy security performance for 30 Chinese provinces, this study chose two types of Multi-criteria Decision Making methods: Fuzzy Analytical Hierarchal Processes (Fuzzy AHP) and Preference Ranking Organization Method for Enrichment Evaluations (PROMETHEE). Fuzzy AHP is frequently used to capture expert knowledge of a preference numerically, and applied to solving hierarchical and complex decision-making problems (Chang, 1996) [85]. Fuzzy AHP uses a membership function to calculate a grade

of membership that a given variable belongs to, and triangular and trapezoidal functions are usually used in fuzzy logic because they are simple to use but also accurate (Ocampo-Duque et al., 2006; Güngör et al., 2009; Heo et al., 2010; Lee et al., 2011; Ren and Sovacool, 2014) [86-90]. The PROMETHEE method is one of the most recent MCDA techniques, firstly developed by Brans (1982) [91]. It has witnessed a rapid proliferation throughout academic research, with several hundred articles published in more than 100 scholarly journals since 1985 (Behzadian, 2010) [92]. 3.2.1 Fuzzy AHP

Our procedure of fuzzy AHP proceed along the following steps:

Step 1: Determining the comparison matrix by using linguistic terms. This step is to make pairwise comparisons of dimensions and metrics by using linguistic terms in the same level of hierarchy structure. In this study, it relies on the linguistic terms and corresponding fuzzy numbers (as presented in Table 2) to make pairwise comparisons.

Step 2: Transforming the linguistic terms into fuzzy numbers. This step aims at transforming the linguistic terms of the pairwise comparisons in Step 1 into fuzzy numbers according to Table 2. Then, the pairwise comparison matrix could be obtained, as presented in Eq. (3)

$$\widetilde{A} = \begin{vmatrix} \widetilde{1} & \widetilde{\alpha}_{12} & \cdots & \widetilde{\alpha}_{1n} \\ \widetilde{\alpha}_{21} & \widetilde{1} & \cdots & \widetilde{\alpha}_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ \widetilde{\alpha}_{n1} & \widetilde{\alpha}_{n2} & \cdots & \widetilde{1} \end{vmatrix} = \begin{vmatrix} \widetilde{1} & \widetilde{\alpha}_{12} & \cdots & \widetilde{\alpha}_{1n} \\ 1/\widetilde{\alpha}_{12} & \widetilde{1} & \cdots & \widetilde{\alpha}_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ 1/\widetilde{\alpha}_{1n} & 1/\widetilde{\alpha}_{2n} & \cdots & \widetilde{1} \end{vmatrix}$$

$$(3)$$

where $\widetilde{a}_{ij} = (a_{ij}^L, a_{ij}^M, a_{ij}^U)$

Step 3: Calculating the fuzzy geometric mean and determining fuzzy weights. The fuzzy geometric mean and fuzzy weight with respect to each factor could be computed by Eqs. (4) and (5), respectively.

$$\widetilde{r}_{i} = (\widetilde{a}_{i1} \times \widetilde{a}_{i2} \cdots \times \widetilde{a}_{in})^{1/n} \tag{4}$$

$$\widetilde{\omega}_j = \widetilde{r}_j / \sum_{j=1}^n \widetilde{r}_j \tag{5}$$

where \widetilde{r}_j is the geometric mean of the fuzzy comparison values of the j_{th} factor to each of the other criteria, and $\widetilde{\omega}_j$ refers to the fuzzy weight of the j_{th} factor.

Step 4: Transforming the fuzzy weights into crisp numbers. The Center of Gravity method was applied to defuzzify the fuzzy weights into crisp numbers, as shown Eq. (6)

$$\omega_j = \left(\omega_j^L + 2\omega_j^M + \omega_j^U\right)/4\tag{6}$$

where ω_j is the crisp weight of the j_{th} factor, $\widetilde{\omega}_j = (\omega_j^L, \omega_j^M, \omega_j^U)$, and $\omega_j^L, \omega_j^M, \omega_j^U$ are the elements of the fuzzy weight.

Our use of PROMETHEE derives from the literature on ranking organization methods for enrichment evaluations (L'Eglise et al., 2001; Macharis and Springael, 2004; Albadvi et al., 2007; Anand and Kodali, 2008) [93-96]. The basic principle of PROMETHEE is based on a pairwise comparison of alternatives along with each recognized criterion, and the alternatives are evaluated according to different criteria, which have been minimized or maximized. More detailed information can be referred from the research of Behzadian (2010) and Brans et al. (1986) [92, 97].

3.2.2 PROMETHEE

The procedure of PROMETHEE is as follows:

Step 1: Computing the deviations based on pairwise comparisons according to Eq. (7)

$$d_{j}(a,b) = g_{j}(a) - g_{j}(b)$$
 (7)

Step 2: Determining the preference function for the criterion. Usually, there are six types of generalized preference function: (1) usual criterion, (2) U-shape criterion, (3) V-shape criterion, (4) level criterion, (5) V-shape with indifference criterion, and (6) Gaussian criterion, as presented

in Eqs. (9)-(14) (Anand and Kodali, 2008) [96]. The more widely used is the linear preference function, which is also known as V-shape criterion.

$$P_i(d) = F_i[d_i(a,b)] \tag{8}$$

$$P(d) = \begin{cases} 1 & d > 0 \\ 0 & d \le 0 \end{cases} \tag{9}$$

$$P(d) = \begin{cases} 1 & d > p \\ 0 & d \le p \end{cases} \tag{10}$$

$$P(d) = \begin{cases} 1 & d > p \\ d / p & d \le p \end{cases} \tag{11}$$

$$P(d) = \begin{cases} 1 & d > p \\ 0.5 & q < d \le p \\ 0 & d \le q \end{cases}$$
 (12)

$$P(d) = \begin{cases} 1 & d > p \\ (d-q)/(p-q) & q < d \le p \\ 0 & d \le q \end{cases}$$
 (13)

$$P(d) = \begin{cases} 1 - e^{(-d^2/2\sigma^2)} & d > 0 \\ 0 & d \le 0 \end{cases}$$
 (14)

where P(d) refers to the preference of alternative a with regard to alternative b on each criteria.

Step 3: Calculating the global preference index according to Eq. (15)

$$\pi(a,b) = \sum_{j=1}^{k} P_j(d) * \omega_j$$
(15)

where $\pi(a,b)$ donates the weighted sum $P_j(d)$ for each criterion.

Step 4: Computing the positive and negative outranking flows with respect to each alternative according to Eqs. (16) and (17).

$$\phi^{+}(a) = \frac{1}{m-1} \sum_{i=1}^{m} \pi(a,i)$$
 (16)

$$\emptyset^{-}(a) = \frac{1}{m-1} \sum_{i=1}^{m} \pi(i, a)$$
 (17)

Step 5: Calculating the net outranking flow with Eq. (18)

$$\mathscr{O}(a) = \mathscr{O}^{+}(a) - \mathscr{O}^{-}(a) \tag{18}$$

There are various kinds of software that were developed to facilitate the PROMETHEE process, including PROMCALC, DECISION LAB and Visual PROMETHEE. In this paper, the Visual PROMETHEE was used to evaluate the energy security performance of China's 30 provinces.

4. Results and analysis

With the dimensions and components of energy security presented in Table 1, the pairwise comparisons with respect to the five dimensions and the components in each dimension were determined as presented in the Appendix A. In order to illustrate how we employed Fuzzy AHP to determine energy security weights, the pairwise comparison with respect to the five dimensions of energy security was firstly transformed into fuzzy scales, as shown in Table 3.

Then, the fuzzy weights of the five dimensions could be determined according to Eq. (4) and (5), and the results were presented as follows:

$$W = (0.2964, 0.2964, 0.1588, 0.1588, 0.0896)$$

In a similar approach, the weight of the metrics in each dimension can also be determined. Then, the global normalized weight of each metric can be obtained, which is the product of the metric's weight in the hierarchy and the weight of its belonging dimension, as presented in Table 4.

Then, we used Visual PROMETHEE to evaluate the energy security of China's 30 provinces for the calendar year 2013.

Table 5 and Figure 1 present our results of provincial energy security performance measurement. The value of Phi is the net outranking flows with respect to the 30 provinces of China, and the Score is the normalized value of Phi according to Eq. (1). The greater the value of Phi or the score, the more secure the energy supply of the corresponding province would be. This data suggests that Shaanxi (0.2413), Shanxi (0.1430), Ningxia (0.1344), Anhui (0.1150), and Qinghai (0.1086) were the five most energy-secure provinces in 2013. All of them are middle and western provinces, rich in energy resources. By contrast, Guangxi (-0.1752), Jilin (-0.1642), Shanghai (-0.1638) Hubei (-0.1601), and Liaoning (-0.1594), performed the worst on energy security metrics for 2013.

According to the varying levels of economic development and geography, the 30 provinces can also be divided into eight regions to provide analysis at a higher scale, namely Northern Coast, Eastern Coast, Southern Coast, Northeast, Middle Reaches of Yellow River, Middle Reaches of Yangtze River, Northwest, and Southwest, regions which are shown in Figure 2. It is apparent that the Middle Reaches of Yellow River performs the best on energy security, since most of China's energy resources locate in this region. This is followed by the Northwest, also a region where energy resources are significant. The Southwest, Southern Coast and Northern Coast are in the third group, and they have achieved moderate energy security performance. However, the Eastern Coast, Middle reaches of Yangtze River, and the Northeast, perform the worst energy security performance.

The next phase of our methodology involved further analyzing the results with PROMETHEE network (see Figure 3). More specifically, the Middle Reaches of Yellow River and the Northwest have the most significant energy reserves and levels of production in China, but these have been

compromised by threats to economic development including weakening infrastructure and environmental degradation. The Southwest, Southern Coast and Northern Coast show a moderate energy security performance. The Eastern Coast, the most economically vibrant area of China, performs poorly due to high levels of consumption, growing dependence on imports, and environmental degradation. Similarly, the Middle Reaches of Yangtze River, lacking fossil fuels and lagging behind the coastal provinces in terms of economic development, performs relatively poorly as well. The worst energy security performer is the Northeast region, whose situation is similar to the Middle Reaches of Yangtze River.

Interestingly, our analysis suggests that the provinces in the Middle Reaches of Yellow River, Northwest, and Northern Coast have shown little variance in their energy security performance. Sichuan, Yunnan and Guizhou, where most of energy resources in Southern China reside, demonstrates relatively high performance. Chongqing and Guangxi show extremely poor energy security performance. The same situation has been observed in the Southern and Eastern Coast regions, where both Hainan and Shanghai show much lower scores of energy security compared with their neighbors. On the contrary, Anhui, rich in coal, has a much better energy security performance than the other three provinces in this region, while Heilongjiang, rich in petroleum, performs better with respect to energy security than Liaoning and Jilin.

The performance score of each dimension for the 30 provinces are shown in Table 6. Grouping the 30 provinces into eight regions, the performance score with respect to each region was determined by calculating the average scores presented in Figure 4.

According to Figure 4, it is apparent that the best energy security performance across all provinces relate to the dimension of affordability and equality, with an average score of 70.40. Chinese provinces demonstrate a moderate performance on technology and efficiency (54.89),

environmental sustainability (51.18), and governance and innovation (55.37). Chinese provinces perform the worst on the dimension of availability and diversity (31.58). In addition, there is rather great variance among the eight regions across three dimensions, namely, availability and diversity which range from 14.93 to 72.35, affordability and equality which range from 39.36 to 91.08, and technology and efficiency which range from 38.27 to 90.21. The other two dimensions, environmental sustainability (ranges from 34.83 to 72.14), and governance and innovation (ranges from 45.20 to 66.53), show smaller variances. Thus, it could be concluded that energy availability and diversity are the main challenges facing China's energy security at the national level.

Moreover, the Middle Reaches of Yellow River and the Northwest show the best performance on availability and diversity. The Eastern Coast, Southern Coast and Northwest are the three regions who perform best on affordability and equality. The Eastern Coast, at the same time, demonstrates a much higher performance on the dimension of technology and efficiency than any other regions. As to the dimension of environmental sustainability, Southern China, including the Southern Coast and Southwest, do better than other regions. As shown above, the eight regions indicate a small variance with each other on the dimension of governance and innovation, but the coastal provinces do perform better than inland provinces.

Lastly, a Sensitivity Analysis was given to the 30 provinces. Firstly, it distributed equal weight to every metric, in which case, a Phi could be obtained for each province, which can indicate their energy security performance. Then, 20 scenarios were proposed, and in the i-th scenario, the weight of the i-th metric equals 0.24, while the weights of the rest metrics equal 0.04. For each scenario, the energy security performance for each province can be computed. Subsequently, the Sensitivity Analysis Factor (SAF) could be calculated, which measures how much a province's energy security performance has change when the weight of a particular metric change from 0.05

to 0.24. After that, the results of Sensitivity Analysis with respect to the eight regions are calculated, as shown in Figure 5. The positive SAF means that the energy security performance is enhanced, while the negative SAF means the performance is weakened. The greater the absolute SAF, the more sensitive the performance is.

Figure 5 also illustrates that the dimension of availability and diversity causes the greatest sensitivity on energy security performance, with an absolute SAF of 0.32, the technology and efficiency, and environmental sustainability lead to a moderate sensitivity, with the absolute SAFs of 0.20 and 0.23, respectively. Energy security performance is least sensitive to the dimension of affordability and equality (absolute SAF is 0.12), and governance and innovation (absolute SAF is 0.13). Specifically, the energy security performance of the Northwest and the Middle Reaches of Yellow River would be greatly enhanced by the increasing the weight of metrics related to availability and diversity, while that of the Eastern Coast and Southern Coast would be severely weakened. The performance of the Eastern Coast and Northern Coast would be improved if the metrics about technology and efficiency were weighted more, but the performance of the Northwest, Southwest and Northeast would suffer. When the metrics about environmental sustainability increase in weight, the Southern Coast, Middle Reaches of Yangtze River and Southwest would benefit while the Northwest, Middle Reaches of Yellow River, and Northern Coast would see performance decline.

5. Conclusions and policy implications

Our results do suggest some salient policy implications for enhancing China's energy security and bridging the gap in energy security performance among different provinces and regions.

Firstly, our data strongly implies that developing renewable sources of electricity and alternative fuels for transport (usually called "New Energy Vehicles" in China) would improve energy security performance and diversify the energy mix. Since energy availability and diversity were determined by our analysis to be the most severe threat facing China's energy security as a whole, national and provincial planners need to invest in more reliable energy sources. China has great renewable energy potential, which can not only maintain its security of energy supply and contribute to diversification, it can also create more jobs and reduce environmental pollution. Coastal provinces in particular, substantial consumers and importers, should find ways to increase such local sources of energy supply. With their comparatively more advanced science and technology sectors, offshore wind power and biofuels perhaps have the most appeal. The Middle Reaches of Yangtze River are gifted with great hydropower and biofuel potential. The Southwest is richest in hydropower as well as unconventional fuels such as shale gas. Each of these resources should be developed to diversify China's energy supply mix.

Secondly, improvements in energy efficiency are needed. On the one hand, most of China's energy resources spatially distant from consumers and located far inland, for instance, coal reserves, existing solar PV deployment, and installed onshore wind power remains in the Middle Reaches of Yellow River and the Northwest; oil and gas reserves are situated in the Northeast and Northwest; hydropower resources are utilized primarily in the Southwest and Middle Reaches of Yangtze River; and biofuel sources are concentrated in Southern China. On the other hand, the inland provinces perform much worse than the coastal provinces on technology and efficiency, and have fewer R&D expenditures and less established infrastructure. Improvements in energy efficiency can optimize the whole life cycle the energy industry, including exploitation, production, processing, transmission and consumption. Thus, R&D on energy efficiency systems such as smart

meters, improved distribution networks to minimize transmission losses, and enhancements to residential, commercial, and industrial building stock is quite beneficial and urgent.

Thirdly, more attention and investment should be paid to mitigating environmental problems and improving environment quality. Though the government has promised that by 2020 China would reduce its carbon intensity by 40 to 45 percent compared to 2005 levels, people remain embroiled in severe environment problems, especially those associated with the mining, processing, combustion, and waste of coal. The government should enforce the adoption of better pollution controls on coal-fired power plants and increase responsible for public infrastructure projects for energy conservation and environment improvement, such as public green spaces and charging stations for electrical vehicles.

Lastly, national Chinese energy strategy should be adjusted to balance development among different regions and provinces. The coastal provinces have strengthened their economies for several decades since the Reform and Opening, and Western China, including the Northwest, Southwest, and a part of Middle Reaches of Yellow River, is under development with the deployment of the Development of the Western Region in China since 2000. However, the Central region of China, including the Middle reaches of Yangtze River, Northeast, and another part of Middle Reaches of Yellow River, have been neglected by national strategy, and show suboptimal energy security performance, a phenomenon termed "Central Sinking" by some scholars (Yang and Zhu, 2007) [98]. The Northeast, who used to be called as the "Eldest Son of the Republic" for its contribution in the planned economy era (OECD, 2006) [99], now also performs poorly on various economic and demographic metrics. Thus, senior Chinese decision makers should focus more on investing in and improving the energy security of the Central Region.

For those outside of China, the study also reveals the complexity of energy security as both a concept and by location. Energy security performance can vary greatly across geographic space. Many provinces in Western China perform much better than the developed coastal provinces. Western provinces, for instance, perform much better in terms of energy availability. More importantly, their relative advantage has been amplified by a smaller population so they can enjoy more energy endowments and resources per capita. The lesson here is relatively simple: Chinese performance on energy security differs greatly within its regions and provinces. Thus, energy analysts should begin to take a far more nuanced approach when they partake in energy security assessments.

Moreover, our study shows that a traditional view of energy security limited to supply security of fossil fuels ignores other salient dimensions such as energy efficiency, energy equality and governance. In fact, the Fuzzy AHP in this study gave higher weights to the dimension of availability and diversity, and affordability and equality, and the 30 provinces demonstrated the greatest variance in the dimension of availability and diversity. Overall, the evaluative framework proposed in this paper consists of five dimensions associated with current elements about energy security. This instrument can be seen as a significant contribution to existing literature on energy security, and it can help evaluate the energy security performance for different regions or countries.

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Appendix A. pairwise comparisons

Appendix A is the pairwise comparisons with respect to the five dimensions and the components in each dimension (see appendix Tables A1-A6).

Appendix B. Supplementary material

Appendix B is associated with the raw energy security data and the process of data analysis.

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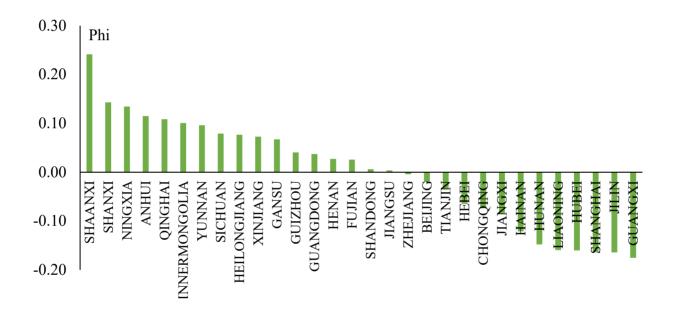
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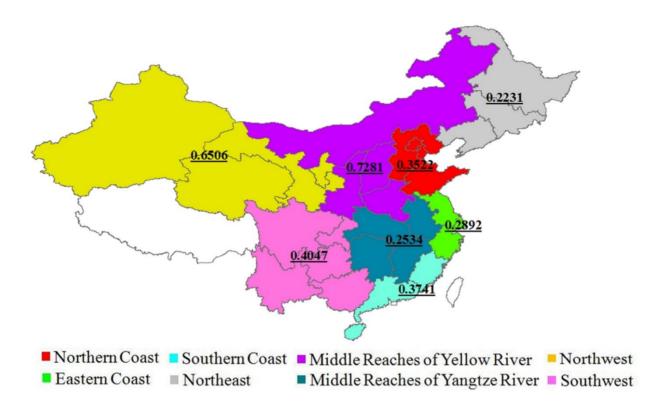
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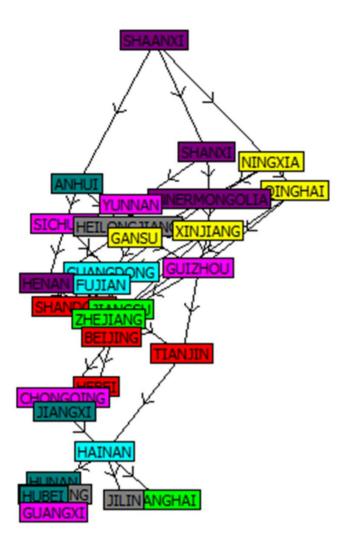
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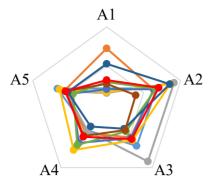
- Figure 1. Net energy security flow ranking 30 Chinese Provinces
- Figure 2. The average energy security performance of eight Chinese regions
- **Figure 3.** The Visual PROMETHEE network of Energy Security Performance for 30 Chinese Provinces
- Figure 4. Energy security performance scores for eight Chinese regions

Figure 5. The result of our Energy Security Performance Sensitivity Analysis









- Northern Coast
- → Middle Reaches of Yellow River
- ---Eastern Coast
- --- Southern Coast
- → Middle Reaches of Yangtze River
- **—**Southwest
- -- Northwest
- -- Northeast
- **→** National

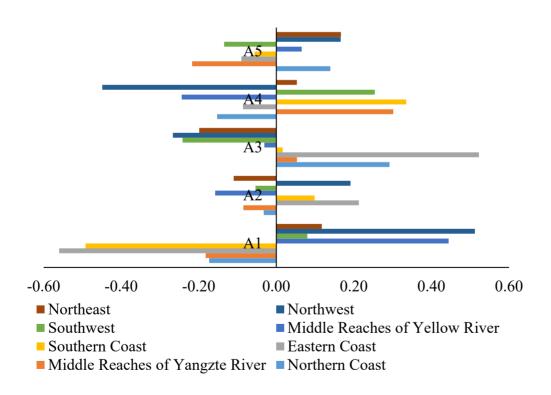


Table 1. Dimensions, components and metrics of energy security

Dimensio	C	Matrice	T.T., :4	D.C.:	Prefere
ns	Components	Metrics	Unit	Definition	nce
$\overline{A_1}$:	A ₁₁ :Security	<i>I</i> ₁ : Primary energy	tce	Comprises the production of coal,	Greater
Availabilit	of supply	production per		crude oil, natural gas and other	prefer
y and		capita		new and renewable energy.	red
diversity	A ₁₂ :Energy	<i>I</i> ₂ : Fossil fuel	tce	Proven recoverable reserves at the	Greater
	potential	reserves per capita		end of a given year divided by	prefer
				total local population.	red
	<i>A</i> ₁₃ :	<i>I</i> ₃ : Self sufficiency	%	Ratio of total primary energy	Greater
	Dependency			production divided by total	prefer
				primary energy consumption.	red
	A14:	<i>I</i> ₄ : Diversity of	-	$SWI = \sum_{i=1}^{m} p_i \ln(p_i)$	Greater
	Diversificati	energy		i=1	prefer
	on	consumption			red
A_2 :	A ₂₁ :Stability	Is: Stability of	-	Standard deviation of gasoline	Greater
Affordabil		gasoline prices		prices divided by the average	prefer
ity and				price.	red
equality	A ₂₂ :Electricit	<i>I</i> ₆ : Coal-fired	Yuan/	Actual prices paid by the	Smaller
	y generation	power tariff	kWh	electricity grid to power plant for	prefer
	cost			the electricity.	red

	A ₂₃ : Electricity equality	I7: Share of electricity in total energy	%	Annul electricity consumed divided by local total energy consumption.	Greater prefer red
	1 7	consumption		1	
	A24:	<i>I</i> ₈ : Quantity of	10^{4}	Local per capita GDP divided by	Greater
	Affordabilit	gasoline bought	tons	the average gasoline prices for a	prefer
	y of	with GDP	per	given year.	red
	gasoline		perso		
			n		
<i>A</i> ₃ :	A ₃₁ : Energy	<i>I</i> ₉ : Energy intensity	tce/10	TPES in tons of standard coal	Smaller
Technolog	efficiency		4	equivalent per 10000 yuan of	prefer
y and			Yuan	GDP.	red
efficiency	<i>A</i> ₃₂ : Grid	I ₁₀ : Electricity	%	Including losses in transmission	Smaller
	efficiency	transmission and		between sources of supply and	prefer
		distribution losses		points of distribution and in the	red
				distribution to consumers.	
	<i>A</i> ₃₃ : Grid	<i>I</i> ₁₁ : Average	Hours	Annual blackout hours per	Smaller
	reliability	blackout hours per		household.	prefer
		household			red
	A ₃₄ : Capacity	I_{12} : Utilization of	%	Annual operation hours of power	Greater
	factor	power plants		plants divided by the total hours	prefer
				for a year.	red

A_4 :	A ₄₁ : Land use	I_{13} : Forest coverage	%	Forest area divided by local land	Greater
Environm				area.	prefer
ental					red
sustainab	A ₄₂ : Water	I_{14} : Waste water	Tons	Annual tons of emitted wasted	Smaller
ility	pollution	emissions per	per	water divided by local	prefer
		capita	perso	population.	red
			n		
	A ₄₃ : Climate	<i>I</i> ₁₅ : GHG emissions	tce	Annual tons of carbon dioxide	Smaller
	change	per unit of GDP	$/10^4 Y$	emissions from fuel combustion	prefer
			uan	divided by local population.	red
	A44:	<i>I</i> ₁₆ : Sulfur dioxide	Tons	Annual tons of sulfur dioxide	Smaller
	Acidificatio	emissions per	per	emissions from fuel combustion	prefer
	n potential	capita	perso	divided by local population.	red
			n		
	A ₄₅ :	<i>I</i> ₁₇ : Nitrogen	Tons	Annual tons of nitrogen dioxide	Smaller
	Photochemi	dioxide emissions	per	emissions from fuel combustion	prefer
	cal potential	per capita	perso	divided by local population.	red
			n		
A5:	A ₅₁ : Market	I_{18} : Investment in	%	Ratio of investment in energy	Greater
Governanc	potential	energy industry		industry divided by total local	prefer
e and				investment.	red

innovatio A ₅₂ :		<i>I</i> ₁₉ : Research	%	Ratio of local government	Greater
n	Innovation	intensity		expenditure on R&D divided by	prefer
	and research			government total expenditures.	red
	A53: Energy	<i>I</i> ₂₀ : Energy saving	%	Ratio of local government	Greater
	&	& environment		expenditure on energy saving &	prefer
	environmen	protection		environment protection divided	red
	t			by government total	
	managemen			expenditures.	
	t				

Source: research interviews, brain storming, and workshop discussion.

Table 2. The linguistic terms and corresponding fuzzy scales

Linguistic scales	Abbreviation	Triangular fuzzy scales
Equally important	Е	(1, 1, 1)
Weakly important	W	(2/3, 1, 3/2)
Fairly strongly important	FS	(3/2, 2, 5/2)
Very strongly important	VS	(5/2, 3, 7/2)
Absolutely important	A	(7/2, 4, 9/2)
Reciprocals of these	RW,RFS,RVS,RA	Reciprocals of the fuzzy numbers

Table 3. Comparison matrix for the five dimensions of energy security

	availability and	affordability and	technology and	environmental	governance and
	diversity (A ₁)	equality (A ₂)	efficiency (A ₃)	sustainability (A ₄)	innovation (A ₅)
A_1	(1, 1, 1)	(2/3, 1, 3/2)	(3/2, 2, 5/2)	(3/2, 2, 5/2)	(5/2, 3, 7/2)
A_2	(2/3, 1, 3/2)	(1, 1, 1)	(3/2, 2, 5/2)	(3/2, 2, 5/2)	(5/2, 3, 7/2)
A_3	(2/5, 1/2, 2/3)	(2/5, 1/2, 2/3)	(1, 1, 1)	(1, 1, 1)	(3/2, 2, 5/2)
A 4	(2/5, 1/2, 2/3)	(2/5, 1/2, 2/3)	(1, 1, 1)	(1, 1, 1)	(3/2, 2, 5/2)
A 5	(2/7, 1/3, 2/5)	(2/7, 1/3, 2/5)	(2/5, 1/2, 2/3)	(2/5, 1/2, 2/3)	(1, 1, 1)

Table 4. Weights of dimensions and metrics for energy security evaluation

Weight of dimensions		Weight of metrics		Global weight	
A_1	0.2964	A ₁₁	0.3315	I_1	0.0983
		A ₁₂	0.3315	I_2	0.0983
		A_{13}	0.1685	I_3	0.0499
		A ₁₄	0.1685	I_4	0.0499
\mathbf{A}_2	0.2964	A ₂₁	0.2095	I_5	0.0621
		A_{22}	0.2481	I_6	0.0735
		A ₂₃	0.2941	I_7	0.0872
		A_{24}	0.2484	I_8	0.0736
A 3	0.1588	A31	0.3408	I_9	0.0541
		A ₃₂	0.2431	I_{10}	0.0386
		A ₃₃	0.2429	I_{11}	0.0386

		A34	0.1732	I_{12}	0.0275
A ₄	0.1588	A ₄₁	0.2491	I_{13}	0.0395
		A ₄₂	0.2490	I_{14}	0.0395
		A43	0.2490	I_{15}	0.0395
		A44	0.1265	I_{16}	0.0201
		A45	0.1265	I_{17}	0.0201
A_5	0.0896	A ₅₁	0.2027	I_{18}	0.0182
		A ₅₂	0.3987	I_{19}	0.0357
		A53	0.3987	I_{20}	0.0357

Table 5. The results of Energy Security Performance for 30 Chinese Provinces

Rank	Province	Phi+	Phi-	Phi	Score
1	SHAANXI	0.4216	0.1803	0.2413	1.0000
2	SHANXI	0.3853	0.2423	0.1430	0.7640
3	NINGXIA	0.3967	0.2623	0.1344	0.7433
4	ANHUI	0.3398	0.2248	0.1150	0.6968
5	QINGHAI	0.3894	0.2808	0.1086	0.6814
6	INNERMONGOLIA	0.3647	0.2639	0.1007	0.6624
7	YUNNAN	0.3434	0.2472	0.0961	0.6514
8	SICHUAN	0.3190	0.2400	0.0790	0.6103
9	HEILONGJIANG	0.3333	0.2565	0.0768	0.6050

10	XINJIANG	0.3505	0.2777	0.0728	0.5954
11	GANSU	0.3300	0.2627	0.0673	0.5822
12	GUIZHOU	0.3320	0.2913	0.0406	0.5181
13	GUANGDONG	0.3093	0.2721	0.0372	0.5100
14	HENAN	0.2887	0.2615	0.0272	0.4860
15	FUJIAN	0.3012	0.2753	0.0258	0.4826
16	SHANDONG	0.2847	0.2786	0.0061	0.4353
17	JIANGSU	0.2949	0.2913	0.0037	0.4295
18	ZHEJIANG	0.2881	0.2923	-0.0042	0.4106
19	BEIJING	0.2808	0.3009	-0.0201	0.3724
20	TIANJIN	0.2901	0.3254	-0.0353	0.3359
21	HEBEI	0.2548	0.3195	-0.0647	0.2653
22	CHONGQING	0.2422	0.3160	-0.0738	0.2435
23	JIANGXI	0.2364	0.3221	-0.0858	0.2146
24	HAINAN	0.2284	0.3495	-0.1211	0.1299
25	HUNAN	0.2028	0.3507	-0.1478	0.0658
26	LIAONING	0.1974	0.3567	-0.1594	0.0379
27	HUBEI	0.1944	0.3544	-0.1601	0.0363
28	SHANGHAI	0.2209	0.3847	-0.1638	0.0274
29	JILIN	0.2116	0.3759	-0.1642	0.0264
30	GUANGXI	0.1891	0.3644	-0.1752	0.0000

 Table 6. Energy Security Performance Scores for 30 Chinese provinces

	Performance Score					
Province	A_1	A_2	A 3	A 4	A_5	
BEIJING	16.65	80.45	66.07	52.14	53.65	
TIANJIN	25.88	62.31	56.63	33.50	60.14	
HEBEI	18.60	63.58	64.14	41.26	70.70	
SHANXI	100	57.92	44.20	39.22	68.82	
INNERMONGOLIA	93.09	64.12	34.47	36.65	60.66	
LIAONING	20.41	45.37	42.88	38.33	48.71	
JILIN	20.14	34.72	34.98	51.57	63.38	
HEILONGJIANG	39.93	37.99	43.48	59.62	55.15	
SHANGHAI	14.05	77.30	89.69	35.93	42.76	
JIANGSU	16.20	95.93	100	36.21	67.38	
ZHEJIANG	14.55	100	80.96	52.55	56.13	
ANHUI	27.71	73.22	66.12	53.94	47.45	
FUJIAN	17.11	86.03	62.26	61.88	55.71	
JIANGXI	16.63	71.16	54.22	67.59	36.99	
SHANDONG	22.43	63.32	75.67	37.38	81.61	
HENAN	23.95	68.74	69.61	45.58	42.81	
HUBEI	16.61	56.99	53.26	52.50	49.99	
HUNAN	18.54	54.59	39.10	69.19	51.37	
GUANGDONG	15.17	97.56	70.32	54.53	100	
GUANGXI	15.30	72.76	37.68	68.46	37.46	

HAINAN	16.12	74.38	37.45	100	35.07
CHONGQING	22.83	58.38	41.45	47.14	64.40
SICHUAN	25.71	64.61	62.52	71.09	44.88
GUIZHOU	48.44	62.78	34.18	53.52	37.85
YUNNAN	32.98	73.95	36.71	80.51	41.43
SHAANXI	72.36	62.95	60.41	58.75	57.34
GANSU	31.47	82.58	38.86	44.1	48.85
QINGHAI	57.52	92.87	28.83	35.51	64.69
NINGXIA	55.09	88.53	48.79	28.25	54.63
XINJIANG	66.14	77.79	36.61	31.44	42.19

Table A1. Comparison matrix with respect to the five dimensions

	availability and	affordability and	technology and	environmental	governance and
	diversity (A ₁)	equality (A ₂)	efficiency (A ₃)	sustainability (A ₄)	innovation (A ₅)
A_1	Е	W	F	F	V
A_2	RW	E	F	F	V
A 3	RF	RF	E	Е	F
A ₄	RF	RF	E	Е	F
A 5	RV	RV	RF	RF	E

Table A2. Comparison matrix with respect to availability and diversity

	Security of	Energy potential	Dependency	Diversification
	supply (A ₁₁)	(A ₁₂)	(A ₁₃)	(A ₁₄)
A ₁₁	E	W	F	F
A ₁₂	RW	Е	F	F
A_{13}	RF	RF	E	W
A ₁₄	RF	RF	RW	E

Table A3. Comparison matrix with respect to affordability and equality

	Stability	Electricity generation	Electricity	Affordability of
	(A_{21})	cost (A ₂₂)	equality (A ₂₃)	gasoline (A ₂₄)
A ₂₁	Е	W	RF	RW
A ₂₂	RW	E	Е	RW
A_{23}	F	E	E	W
A ₂₄	W	W	RW	E

Table A4. Comparison matrix with respect to technology and efficiency

	Energy	Grid efficiency	Grid reliability	Capacity
	efficiency (A ₃₁)	(A_{22})	(A ₃₃)	factor (A ₃₄)
A31	Е	W	F	F
A ₃₂	RW	E	W	RW

A33	RF	RW	E	F
A34	RF	W	RF	E

Table A5. Comparison matrix with respect to environmental sustainability

_		Land use	Water	Climate	Acidification	Photochemical
		(A ₄₁)	pollution (A ₄₂)	change (A ₄₃)	potential (A ₄₄)	potential (A ₄₅)
_	A41	Е	RW	RW	F	F
	A ₄₂	W	E	Е	F	F
	A43	W	E	E	F	F
	A44	RF	RF	RF	E	W
	A45	RF	RF	RF	RW	E

Table A6. Comparison matrix with respect to governance and innovation

	Market	Innovation and	Energy & environment
	potential (A ₅₁)	research (A ₅₂)	management (A ₅₃)
A51	Е	RF	RF
A ₅₂	F	E	W
A ₅₃	F	RW	E