

# Methodology to detect the climate change impact on national stability

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## Abstract

This study explored the influence of climate factors on national stability by establishing a hierarchical model based on FSI indicators and correlation test. In the research process, we used data from nine countries around the world to verify the effectiveness of the model, and based on this, we identified the impact of climate factors: temperature, precipitation, and climate disaster index on national stability. Then, in case studies of Yemen, Germany and Egypt, we analyzed the impact of climate change on their stability based on the model we built and proposed interpretations or advise, respectively.

**Keywords:** National Stability Index (NSI); Pearson correlation test; Moving Average Method

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# 1 Introduction

## 1.1 Background

Climate change has been an unneglectable issue identified globally in 21th century. Historically, the evolution of climate change was triggered by factors such as biotic processes, variations in solar radiation received by Earth, plate tectonics, and volcanic eruptions. Recently, human activities have also been identified generally as primary causes of ongoing climate change, often referred to as global warming.

Along with the exacerbation in global climate, the living environment as well as the natural resources degenerates in unequal extent to different region of the world, which is susceptible to increasing the fragility of nations. Countries like Rwanda, Africa, are likely to become the victims for such process. Droughts of varying intensity are common in sub-Saharan Africa, and the continent has suffered eight serious droughts since World War II. The pressure brought by the increasing population against the decreasing arable land areas and productivity tensed the interaction between the tribes inside the country. This instability eventually deteriorates to an appalling genocide launched by the Hutu led government towards the Tutsi members in 1994.

Though lack of explicit clues to verify the casual relationships between the climate change and national fragility, the worsening climatic and environmental conditions in those countries with severe social unrest still shed light to us on the potential linkage of climate change and country fragility. In this paper, we try to analysis the impact of climate anomalies based on a multi-layer hierarchical model, which will be elaborated in section two.

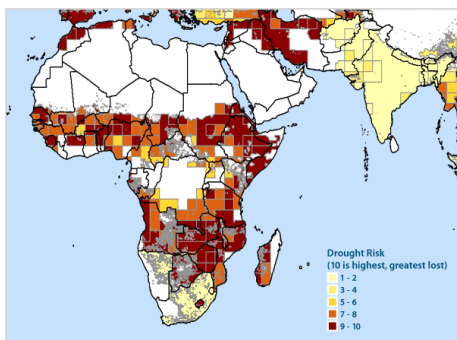


Figure 1: Natural Disasters Hotspots

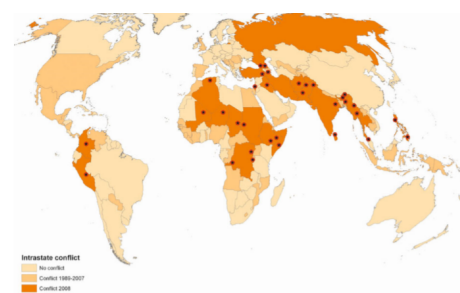


Figure 2: Contemporary Armed Conflicts in 2008

## 1.2 Literature Review

The previous investigators provided some fundamental ideas for the fragility of states. Basically, a fragile state is always a low low-income country characterized by weak state capacity and/or weak state legitimacy leaving citizens vulnerable to a range of shocks. Generally, fragile states include(*Wikipedia*):

- post-conflict/crisis or political transition situations
- deteriorating governance environments
- situations of gradual improvement and
- situations of prolonged crisis or impasse

Concerning researches conducted on the associations between climate change and social stabilities, few of qualitative methods are put forward or validated. Amy Richmond proposed a regression model to analyze the relationship between political stability with non-violence (PSV) and arable land as well as the ratio of population having access to improved water. The result shows a significant positive correlation between the dependent variable and independent variables, which means countries with more arable land and access to improved water will perform better in political stability with non-violence. (*Modeling Environmental Security in Sub-Saharan Africa*, Amy Richmond Krakowka, 2012)

A thorough collection of the discussion of the challenges for society stability associated with climate change is *Climate Change, Human Security and Violent Conflict* edited by Jürgen Scheffran, et al. This literature illustrates the influence attributed to climate change on security, securitization discourse and migration as well as some cooperative policies and capacity-building approaches to improve the climate condition. Some frameworks and modelling theories to study the linkage between climate factors and states stability are also presented.

## 2 Hierarchical Model

### 2.1 Reference Model

In the second section, the methodology and structure of our national stability evaluation model will be demonstrated. The hierarchy of our model refers to the structure of fragility states index (FSI). FSI is an annual ranking of 178 countries based on the different pressures they face that impact their levels of fragility. The assessment method of this system is based on The Fund for Peaces proprietary Conflict Assessment System Tool (CAST), an analytical approach, which requires three primary streams of data: quantitative, qualitative, and expert validation.

The data are triangulated and subjected to critical review to obtain final scores for the FSI.

The model assigns scores to the states from four aspects: cohesion, economy, political and society, each representing a mediate indicator. Every indicator consists of three root indicators as Figure 3 suggests. The altogether 12 root indicators are the final scoring indexes, and the original score for each root indicator ranges from 0 to 10. The rank is made by the total summing up score of the root indicators. Higher score will be interpreted as a sign of greater fragility.



Figure 3: FSI Model Indicators

## 2.2 Assumptions and Annotations

In order to successfully extract the essential elements in the relationship of climate and country fragility and build a general model, we made the following assumptions to simplify the framework of the model:

- The pattern that meteorological factors influence the stability of each country is set to be undifferentiated regardless of the individuality of the countries. That is to say, the factors that are identified to account for the variance in fragility are fixed for all the sample states. When expressed in mathematical forms, the function structures of the states are assumed to be same. The difference of the countries is expressed only through the discrepancy of the coefficients of the functions, though we admit the existing possibilities of diverse influence patterns for countries in different areas.
- The pattern that meteorological factors influence the level of fragility of a country over time is also assumed to be of constant structures. The initiative of this assumption is coherent to that of assumption one, and the time-related change is also indicated by variance of the coefficients.

The following notations notify the components of the multi-layer model we defined. The indicators in the bottom layers, whose scores are derived from existing external evaluation indicators, are of the third layer, while others of the second and first layer are indicators defined by us, whose values are derived from those in the third layer.

First-layer indicator	Notation	Second-layer indicator	Notation	Third-layer indicator	Notation	Third-layer indicator	Notation
Political stability index	P	Administration Index	ADI	Institutions and Infrastructure	II	Talent attract	GTCIA
Social stability index	S	Financial Index	FNI	Democracy index	DI	Talent growth	GTCIG
Economical stability index	E	Military Index	MLI	Crime Index	CI	Talent remain	GTCIR
Natural Environmental stability index	N	Equality Index	EQI	Financial deficit indicator	FD	Population density	PD
		Welfare Index	WFI	Global Fire Power	GFP	cultivated area	CA
		Security Index	SCI	Gini coefficient	GINI	Cereal production	CP
		Democracy Index	DEI	Health & education & training	HET	Per capita renewable freshwater resources	PCRF
		Macro Economy Index	MEI	Global Terrorism Index	GTI	Forest area	FA
		Employment Environment Index	EEI	Number of refugees	NOR		
		Population Pressure Index	PPI	GDP	GDP		
		Natural Resource Index	NRI	Unemployment Rate	UR		

Figure 4: Indicator Notation

## 2.3 Model Structure and Methodology

Our multi-layer model of national stability has a similar hierarchy to that of FSI model. We extended the reference model to a three layered one while replaced several indicators in FSI with other indexes which are more suitable to analyze the climatic impact (e.g., cultivated area, renewable freshwater resources and forest area). The visualized structure of our model is presented in Figure 5:

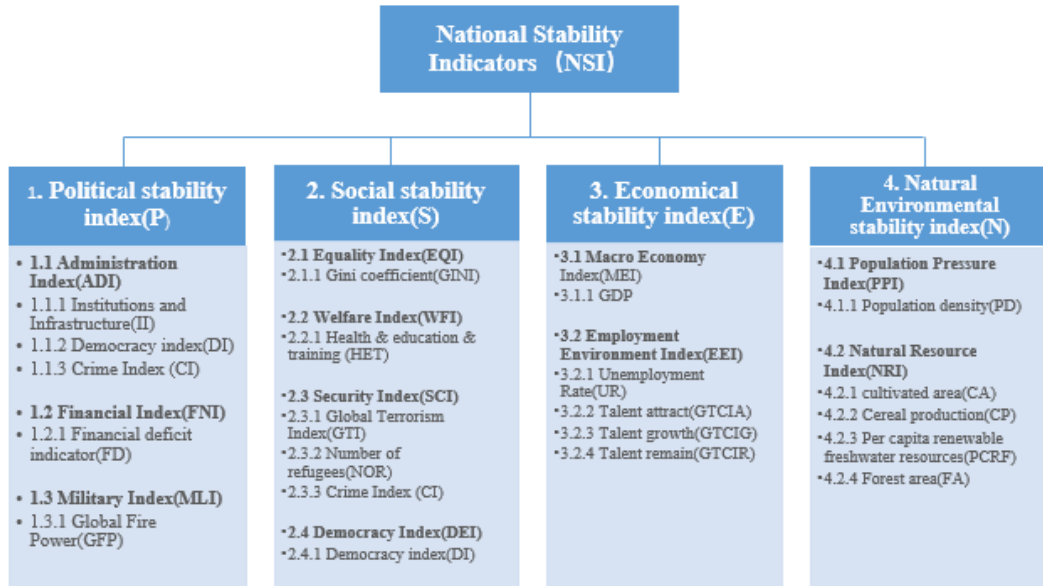


Figure 5: Multi-layer (NSI) Model Structure

The third layer of the model consists of widely acknowledged indexes like GDP and GINI coefficient (measuring the gap between) as well as objective indexes such as cultivated area and renewable freshwater resources. The selected indexes are highly relevant to the fragility of the nations in accordance with the FSI-

*methodology.* The indicators in first and second layer are produced by the same logic, being clustered from secondary indicators (i.e. the second layer indicators are calculated from its secondary indicators in the third layer) based on empirical rules and reference to the FSI indicators (see in Appendix. Figure 15). In addition, an original first layer dimension in FSI, cohesion, is replaced because of great overlap it shows with other dimensions. Instead, we added the dimension of natural environment into this layer due to the idea that environment is easily affected by climate change while also takes a explicit place in state stability.

The model can function as a ranking system like FSI model, but the difference is that we measure the opposite side of state fragility, the national stability. Thus, the model can hereof be named as NSI model. The ranking methodology of this model is through assigning scores to the countries to be evaluated, where higher score implies higher level of stability in the country. The approach of the scoring is from the bottom (third layer) to the top, which means the first step is to handle the external data from the third layer indicators to obtain the values for second layer indicators. The mathematical expressions for this step are:

$$ADI = \frac{II + DI + CI}{3} \quad (1)$$

$$FNI = FD \quad (2)$$

$$MLI = GFP \quad (3)$$

$$EQI = GINI \quad (4)$$

$$WFI = HET \quad (5)$$

$$SCI = \frac{GTI + CI + NOR}{3} \quad (6)$$

$$DEI = DI \quad (7)$$

$$MEI = GDP \quad (8)$$

$$EEI = \frac{UR + \frac{GTCIA+GTCIG}{2} + GTCI}{3} \quad (9)$$

$$PPI = PD \quad (10)$$

$$NRI = \frac{CA + CP + PCRf + FA}{4} \quad (11)$$

The basic logic for this step is to score the second layer indicators by gaining the average value of subordinate indicators in the third layers. Noticing that in equation (9), we combine the indicators of GTCIA and GTCIG into one index since they both reflect the increase of talents in a state. The procedure of scoring the first layer indicators and final NSI is in consistence with this. The specific

equations are listed below:

$$P = \frac{ADI + FNI + MLI}{3} \quad (12)$$

$$S = \frac{EQI + WFI + SCI + DEI}{4} \quad (13)$$

$$E = \frac{MEI + EEI}{2} \quad (14)$$

$$N = \frac{PPI + NRI}{2} \quad (15)$$

$$NSI = \left( \frac{P + S + E + N}{4} \right) \times 100 \quad (16)$$

Additionally, data preprocessing is necessary for eliminating the effect of the scale of the data. We standardize and normalize the data by dividing them with the sum of all values in the same index, in order to constraint the input data into [0,1]. Some index like UR is actually negative correlated to its parental indicator. In this case, we conduct linear transformation to the value of these indexes to make them positively correlate to the parental indicators, which suits our scoring criteria.

After the scoring and ranking process, we are able to assess the level of stability of a nation. We are required to categorize the countries in three states: stable, vulnerable and fragile. The assessment benchmark is anchored in the concept of long tail theory, which indicates that about 20% of the population accounts for 80% of the social wealth. When this theory is applied to the study of national stability, most countries should be viewed as stable or vulnerable states, and only a few ones that have assembled the most destabilizing factors, should be labeled as fragile states. Hence, we set up the rules for classification as Figure 6 suggests:

Assessment	FSI Rank	FSI Score	NSI Score
<b>Fragile</b>	1-10	>106	<7
<b>Vulnerable</b>	11-60	82-106	7-10
<b>Stable</b>	61-180	<82	>10

Figure 6: Assessment Benchmark

## 2.4 Model validation

The method of model validation is based on the Pearson correlation analysis of final stability (fragility) scores in our model and the reference model (see Appendix Table 6). If the correlation test shows significant correlations between our model and FSI, then the effectiveness of our model is verified. We did the test based on a sample containing relevant data of 9 countries in 2014. The test results from are displayed in Figure 7 and Figure 8.

Correlations			
		NSI	FSI
NSI	Pearson Correlation	1	-.741*
	Sig. (2-tailed)		.022
	N	9	9
FSI	Pearson Correlation	-.741*	1
	Sig. (2-tailed)	.022	
	N	9	9

\*. Correlation is significant at the 0.05 level (2-tailed).

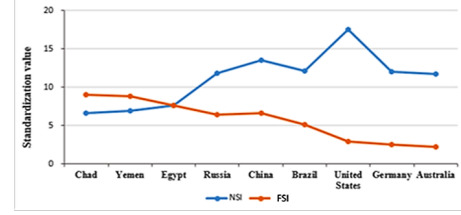


Figure 7: FSI & NSI Correlation Table      Figure 8: Scoring Trend Comparison

The result indicates that the score of our model has strongly negative correlation to the FSI model, which is consistent with our expectations. Therefore, the model is qualified to identify the stability of the countries.

## 2.5 Climatic Factor Impact Mechanism

The impact of the climatic factors on the stability (fragility) is divided into two patterns, the direct pattern and indirect pattern. In our study, the direct pattern is interpreted as a strong correlation between the climatic factors and the final NSI scores, while indirect impact is defined as climatic factors posing impact through the subordinate indicators of the model, such as those in first and second layers of the model.

The climatic factors we chose incorporates three dimensions: temperature, precipitation and climatic disasters. The disaster dimension is measured with the natural disaster risk, while temperature is measured by annual average temperature and precipitation is calculated by the annual rainfall depth with climatic data inquired from NASA and World Bank. The correlation test shows that the climatic factors affect the stability in indirect pattern as Figure 9 suggests, instead of in a direct way (see Appendix. Figure 17). The specific approach is that temperature functions on NRI and MLI while precipitation affects NRI and natural disasters impact on MLI, WFI and EEI. (see Appendix. Figure 18)

The Pearson correlation test indicates that the climate factors defined in our model are likely to possess the linear correlations to the indicators. We produce the mathematical functions in linear format to explicitly illustrate the impact patterns:

$$Climate\ factors = [t, p, d]$$

$$MLI = A_{ij} \times t + B_{ij} \times d + \alpha_{ij}$$

$$NRI = C_{ij} \times t + D_{ij} \times p + \beta_{ij}$$

$$WFI = E_{ij} \times d + \gamma_{ij}$$

$$WFI = F_{ij} \times d + \delta_{ij}$$

(17)

$t$  – temperature     $p$  – precipitation     $d$  – disaster     $i$  – indice of countries

$j$  – indice of time serials     $X_{ij}$  – coefficient of climate factor impact



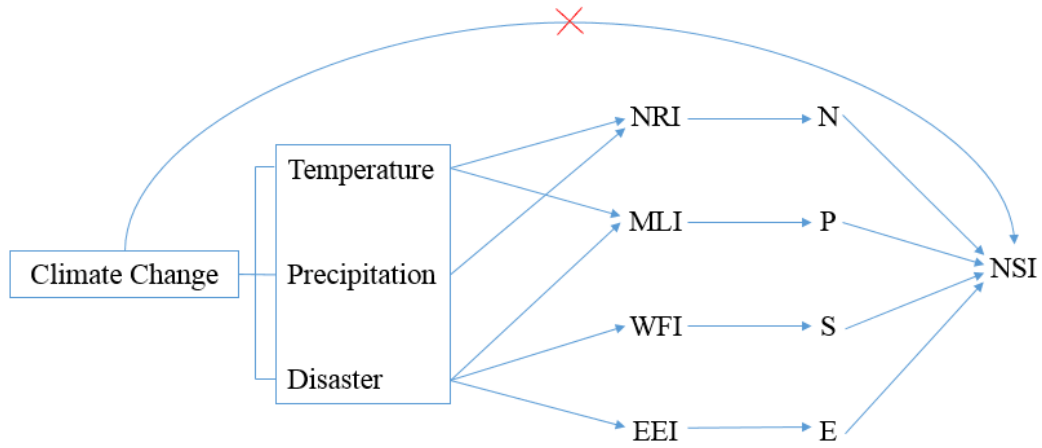


Figure 9: Impact Mechanism Chart

### 3 Case Study

#### 3.1 Analysis of Yemen-Controlled Analysis of Climatic Factors

Yemen is a country located in the southwest of the Arabian Peninsula, where the amount of precipitation is relatively small compared to the global average level. In FSI system, Yemen is listed at the fourth place which indicates a comparatively higher fragility than most other states. In our sample, Yemen is categorized as a fragile country with the total NSI of 6.95. These signals provide us with insight to look into the impact that climate factors might impose on Yemen.

The analysis was progressed by a controlled test. The experimental group is an adjusted NSI model, whose second layer indicators influenced directly by the climate factors are removed. Meanwhile, the comparative group is the ordinary NSI model. That is to say, the adjusted NSI model does not include the second layer indicators of MLI, NRI, WFI and EEI. Thus, the equations for the first layer indicators are changed into following format:

$$P = \frac{ADI + FNI}{2}$$

$$S = \frac{EQI + SCI + DEI}{3}$$

$$E = MEI$$

$$N = PPI$$

After such adjustment, the NSI score for Yemen increased from 6.949 to 7.485 (increase of 7.32%) as the graph shows. From the four dimensions in the first layer, we can see that in political, social stability and natural environmental indicators, the elimination of climatic factors has improved value performance more or less.

Particularly, the natural environmental index obtained an obvious enhancement. The subordinate indicator for this index is NRI, which accounts for the resources in the country and impacted by precipitation and temperature. This is consistent to the prior experience mentioned above. The scarcity in precipitation has reduced the amount of the natural resources in Yemen, which in turn exacerbate the fragility of the nation. Surprisingly, the economical index decreased at the absence of climate factors. In the dimension of economic stability, the correlated index is EEI. It is probable that the contribution made by other subordinate elements (e.g. GTCIA and GTCIG) of EEI largely overweigh the limitation formed by climate conditions making the removal of EEI a loss in economic score overall. But the existence of climate factors lessens the stability of Yemen, conclusively.

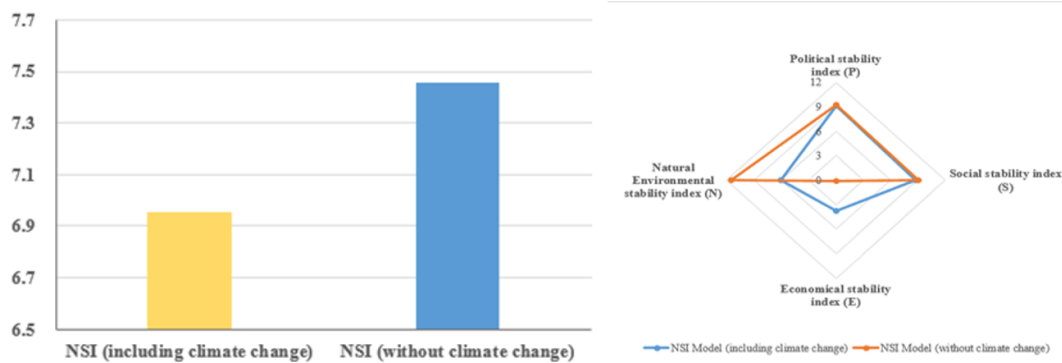


Figure 10: Scores Comparison of Yemen before and after the separation of climate factors

### 3.2 Analysis of Germany-Time Series Analysis

Germany is located in the middle of Western Europe, with abundant precipitation and suitable temperature. In the FSI list, Germany is in the 165th place, and in our NSI index evaluation system, it is also a country of stability. This is in line with the fact that it is labelled as a developed country. However, as the deterioration of the global climate, it is reasonable to worry about that stable states like Germany will fall into a chaotic situation someday. Therefore, we choose Germany to conduct the time series prediction analysis.

The method for prediction of the fluctuation of climate factors over time is realized by applying simple moving average method. This method estimates the future value of the objects by averaging the values in nearest time-point. We use the data of Germany from 2011 to 2015 as the basis for moving. The step length is self-defined, and we set it to be two since this value could lead to a maximized sensitivity towards past data. In order to precisely measure the impact of climate factors, we set the other scoring indexes to be fixed in the simulation.

Based on the climate factors impact pattern we detected, we also compare the variation in the impact of the meteorological factors over time by contrasting the climate factors coefficients in period 2011-2015 to those in period 2011-2045

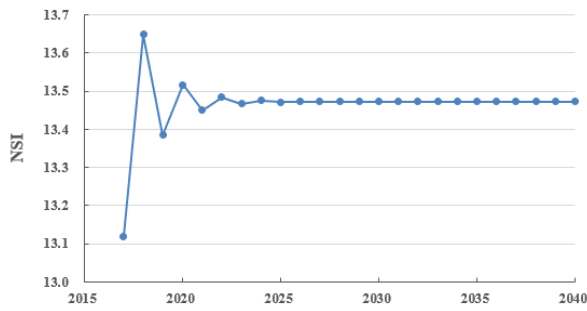


Figure 11: Simulation Chart

Year	NSI	Year	NSI
2017	13.120	2028	13.473
2018	13.649	2029	13.472
2019	13.384	2030	13.472
2020	13.517	2031	13.472
2021	13.450	2032	13.472
2022	13.483	2033	13.472
2023	13.467	2034	13.472
2024	13.475	2035	13.472
2025	13.471	2036	13.472
2026	13.473	2037	13.472
2027	13.472	2038	13.472

Figure 12: Predicted Time Serials Data

(predicted). From Figure 13, while most coefficients stay in positive state, we can find out that the coefficient D and F switch from positive to negative, indicating that there might be a variance in the specific trend of factors impacts. Taking the coefficient F as an example, F measures the correlation between natural disaster risk and welfare index, and the coefficient is positive in the period of 2011-2015, indicating that the occurrence of natural disasters might prompt the German government to sow more subsidies or other forms of social security to improve people's situation in the face of natural disasters. In the whole forecast period from 2011 to 2045, the coefficient is negative, possibly implies that in the later stage of the forecast period, the increase of disasters will make the government more vulnerable and unable to continue to bear the high welfare policies.

		A	B	C	D	E	F
2011-2015	coefficient	0.002475	0.773344	0.000690	0.000014	0.640953	0.692704
	std	0.003195	1.382909	0.000445	0.000008	0.290564	2.880074
2011-2045	coefficient	0.005010	0.507085	0.063764	-0.000773	0.591862	-1.135084
	std	0.001896	0.841814	0.027230	0.000510	0.074496	0.874938

Figure 13: Germany's Climatic Coefficient Table

The figure implies that if the climate factors varies in the same pattern as in 2015 and the impact level for the basic indicators remain the same, Germany is unlikely to become unstable by 2040. On the opposite, its NSI scores might experience an increase and finally reach a higher level. This simulation result indicates that the impact brought by climate factors is majorly positive at present. This could be verified through the positivity of the coefficients. Another explanation is that the extent of the impact is pretty small since the coefficients are not large in absolute value.

According to the above statistics, if the climate factors varies in the same pattern as in 2011 to 2015, Germany is unlikely to become unstable by 2040. On the opposite, its NSI scores might even experience an increase and finally reach a higher level. This simulation result indicates that the impact brought by climate factors is majorly positive at present. This could be verified through the positivity-

ty of the coefficients. However, in this analysis we neglect the possibility of great variance in the coefficients as well as that of the patterns of the climate change, which cannot be obtained through past data. When those coefficients become negative and large enough in absolute value, which is always associated with a terrible climate condition, the stability of Germany might be swung.

### **3.3 Analysis of Egypt-Estimation of Human Interventions**

Egypt is located in northeast Africa, where most of the country is in a dry and hot condition with growing population. Based on these traits, it is suitable for us to target Egypt for research on human intervention to reduce the impact of adverse weather conditions. We brought Egypt's 2017 specific data indicators into the model to calculate their stability score, and then compare it with the previous value. The result shows that it was in the vulnerable level. In task 1, we have already known that the indicators related to climate change are military strength (MLI), social welfare (WFI), employment environment (EEI) and natural resources (NRI). According to these four indicators, the government could take relevant measures to enhance the stability of the country, that is, raising the NSI value, from the aspect of climate conditions. We will propose specific measures for the above indicators to mitigate the impact of harmful climate change on the vulnerability and estimate the cost of these interventions.

In terms of natural resources (NRI), Egypt is an extremely water-scarce country. Per capita availability of fresh water, forest area, grain yield and other indicators are far below the world average. The shortage of natural resources such as water resources has increased the vulnerability of the country. Therefore, in response to this issue, we try to firstly improve the NSI score of Egypt at the angle of NRI, especially making up for its lack of natural precipitation. We can take methods to reduce the impact of climate change on natural resources like improving sanitation water coverage in urban and rural areas, sewage management, optimizing water resource usage by improving irrigation efficiency and agricultural drainage reuse, building a desalination project base to make up for shortages of water and so on. Among these approaches, we hereof highlight the construction of desalination projects. In 2017, the Egyptian government plans to start construction of 16 seawater desalination plants in five provinces along the coast. After the completion of the project, the daily output of these desalination plants will reach 473,000 cubic meters. Egypt currently has 39 seawater desalination plants, daily freshwater production of 235,600 cubic meters.

We assume that the cost of investment is 1 billion US dollars, and the variable cost of the project depends on how desalination is done. In our study, the way of desalination is reverse osmosis, costing about 0.8 US dollars / ton, with a desalination rate about 60%. This measure can make per capita renewable water (PCRF) increased, as a result the NSI increased. Then we will estimate the total cost of this method.

$$\begin{aligned}
Totalcost &= \text{fixed cost of investment} + \text{variable cost} \\
&= \text{fixed cost of investment} + \text{unit cost} \times \text{amount of desalination} \\
&= \text{fixed cost of investment} + \text{unit cost} \times \text{annual output} / \text{desalination rate}
\end{aligned} \tag{18}$$

In terms of social welfare policies, the government can raise the level of medical insurance and increase the guarantee for natural disasters, so as to reduce the quality of life caused by climate change and other factors. We assume that the investment of social security system can increase the WFI by 13% with one billion dollars. Combined with investment on desalination and social security, we can calculate the minimum annual freshwater output needed to be increased to lift Egypt to a stable state in ten years. The calculation is based on the following equation:

$$\begin{aligned}
\Delta NSI &= 100 \times (0.25 \times 0.25) \Delta WFI + 100 \times (0.25 \times 0.25 \times 0.5) \Delta PCRF \\
&= 6.25 \times 0.13 \times WFI + 3.125 \times \Delta PCRF
\end{aligned} \tag{19}$$

Given the original WFI and NSI, the result for additional minimum annual freshwater output is about 0.74 per capita. Considering the approximate total population of one hundred million, the overall freshwater needed could be seventy-four million cubic meters. The predicted NSI is shown in Figure 14:

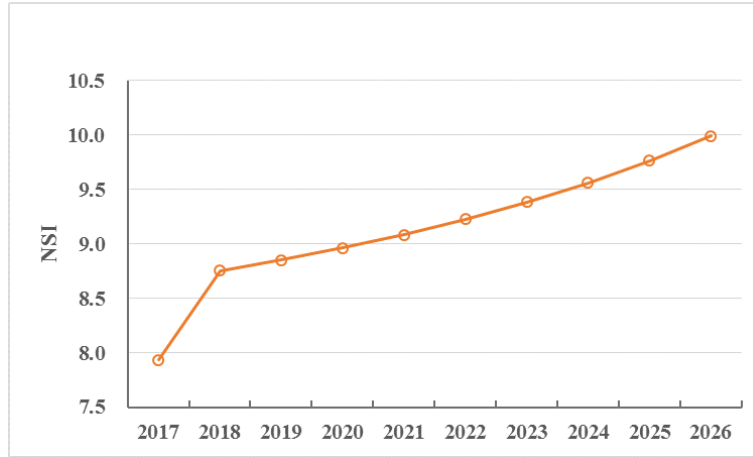


Figure 14: Egypt's Trend Prediction

However, these improvements require over two billion dollars according to our assumptions. A trade-off is likely to be made between the huge cost and the goal to achieve stability, though the cost might be varied under other conditions. Governments can smooth the burden of investment by reducing the intensity of one-time investments and lengthening the investment cycle.

Besides, Egypt's unemployment rate and talent competition index are also well below average. In terms of the employment environment (EEI), the government need to provide more jobs for unemployed people caused by adverse climate conditions and other reasons. The government also needs to strengthen its training of qualified personnel and enhance the index of talent competitiveness so as to raise the economic indicators and the degree of soundness of the country.

### **3.4 Possible Application to States/Continents**

In our opinion, the applicability of our model largely depends on whether the necessary data is attainable. The NSI model is based on the country's composite score on various indicators to determine the country's stability. The indicators for each country are scored from well-known reports or indexes, which can be obtained through multiple publications and websites (*e.g. World Bank, Wikipedia*). In the judgment of continents or cities, the indicators and criteria we need to consider are roughly similar.

For larger regions, such as continents, if some necessary data is not provided, the NSI scores can be obtained by averaging or weighted averaging the country data for that continent. For example, the GDP indicator can be used in evaluating the economic strength of a country, and we can also use the average GDP of Asian countries to assign value for the scoring of NSI of Asia. However, for smaller regions, such as cities, the lack of data is difficult to make up for, which will increase the difficulty of using our model. In the case where the accuracy requirement is not high, the NSI score of the country can be used to approximate the stability score of the city. The best solution is to get all the index data you need for scoring.

Besides the model itself, we want to highlight the analytical methodologies. For different regions, or larger and smaller regions, the influence pattern of climate factors can be varied. Further researchers can test the correlation between the climate factors they choose and the NSI index, so as to determine the specific impact model of climate factors and the indicators directly affected. We believe that this identification approach should not be limited by the scale of the analysis object (*i.e.*, country, city or continent).

## **4 Strengths and Weaknesses**

One of the main advantages of this model is that its indexes is no single, describing multiple dimensions and indexes for researches to detect climate change impact on national stability. The model expression is relatively simple and easy to be comprehended, the calculation is relatively easy and the data is more convenient to obtain, which can be found in the relevant reports. The data we calculated through the model have strong correlation with the FSI index, which shows

that our model is effective. Through the calculation of each index, we can not only get the NSI index, but also obtain the index value of the stability condition on the specific dimension (i.e., political, social, economic and environmental) of the country., which provide further insights for policy-making to deal with undesirable climate changes.

In addition to the model, the identification method we used in this study, which focus on the Pearson Correlation test also provides more flexibility for the use of the model. Subsequent researchers can conduct correlation tests between the climate factors determined according to their own needs and indicators in NSI to identify the impact mode of climate, rather than being limited to the temperature, precipitation and disaster index selected in this project.

However, the defects of the model also exist, mainly from the model hypothesis, linear relationship hypothesis, subjectivity of index selection and similarity between indexes. In the model hypothesis, we assume that the impact structure is fixed among different countries and periods, which might not hold true in reality. The linear relationship hypothesis refers to the fact that in this model, we limit the conversion between different indexes in different layers and the correlation between climate factors and indicators to a linear form. Other possibilities, such as exponential relationships, were not considered in this study. In terms of the selection of model indicators, although we refer to the FSI model and conduct a certain correlation test, subjectivity cannot be completely eliminated when indicators are deleted or added. Finally, because some indicators themselves are converted by other indexes, there will be some overlap or collinear between the indicators, which reduces the value of regression analysis and the internal distinction of the model.

## 5 Conclusion

In this study, we conducted a modeling analysis on the impact of meteorological factors on national stability and successfully identified the mechanism by which meteorological factors affect national fragility (stability). The main contributions of this paper are as follows:

1. We successfully built the model on the relatively well-established FSI system. On the two-layer structure of the reference model, we refine the model to three levels and reevaluate important indicators to construct an NSI model that is more suitable for analyzing climate impacts. The model can give the corresponding stability scores to the analysis object, and we combine the long tail theory to make the scoring results fragile, vulnerable and stable state partition. The validity of the model passed the Pearson correlation test, which can also be basically in line with the rule of thumb in the case analysis.
2. Based on the self-established NSI model, we use the Pearson correlation coefficient test to complete the pattern analysis of the impact of meteorological factors

on the national stability. It is found that the meteorological factors affect the national stability indirectly by acting on the second layer of indicators rather than directly affecting the national stability index.

3. Combined with the control experiment, the moving average method and so on, we analyze or predicts the specific influence of the sample countries by climate factors and puts forward some insights on the dealing of these influences.

## 6 Further work

In the follow-up process, we will collect a large amount of data for model matching and testing, so as to select better indicators and continuously improve the expression and effectiveness of the model. In analyzing the impact of climate change, we have selected only a few representative indicators as climate factors. We would like to collect more data about climate characteristics in order to describe the impact of change in different climate factors on the national vulnerability influences in a more comprehensive way.

Besides, we would like to explore the potential individuality we avoid in this general model, to extend our model to suit diversified conditions. Also, we have interested in exploring the climate impact pattern in other function format other than linear, such as exponential etc.

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## Appendices

	SA	FE	GG	EC	UD	HF	SL	PS	HR	DP	RD
ADI	*	*	*				*	*	*		
FNI				*							
MLI	*						*				
EQI		*			*						
WFI					*			*		*	
SCI	*							*		*	*
DEI		*	*				*		*		
MEI				*							
EEI				*	*	*					
PPI										*	
NRI										*	

Figure 15: Indicators Correlation Between NSI and FSI

Country	NSI Score	FSI Score
Chad	6.6235	108.7
Yemen	6.9493	105.4
Egypt	7.6087	91
Australia	11.7586	26.3
Russia	11.857	76.5
Germany	12.0084	30.6
Brazil	12.1566	61.4
China	13.5252	79
America	17.5126	35.4

Figure 16: NSI versus FSI Scores for Nine Sample States

Correlations <sup>o</sup>		NSI <sup>o</sup>	t <sup>o</sup>	p <sup>o</sup>	d <sup>o</sup>
NSI <sup>o</sup>	Pearson Correlation <sup>o</sup>	1 <sup>o</sup>	-.607 <sup>o</sup>	.458 <sup>o</sup>	-.387 <sup>o</sup>
	Sig. (2-tailed) <sup>o</sup>	<sup>o</sup>	.083 <sup>o</sup>	.215 <sup>o</sup>	.304 <sup>o</sup>
	N <sup>o</sup>	9 <sup>o</sup>	9 <sup>o</sup>	9 <sup>o</sup>	9 <sup>o</sup>
t <sup>o</sup>	Pearson Correlation <sup>o</sup>	-.607 <sup>o</sup>	1 <sup>o</sup>	.050 <sup>o</sup>	.298 <sup>o</sup>
	Sig. (2-tailed) <sup>o</sup>	.083 <sup>o</sup>	<sup>o</sup>	.898 <sup>o</sup>	.436 <sup>o</sup>
	N <sup>o</sup>	9 <sup>o</sup>	9 <sup>o</sup>	9 <sup>o</sup>	9 <sup>o</sup>
p <sup>o</sup>	Pearson Correlation <sup>o</sup>	.458 <sup>o</sup>	.050 <sup>o</sup>	1 <sup>o</sup>	-.119 <sup>o</sup>
	Sig. (2-tailed) <sup>o</sup>	.215 <sup>o</sup>	.898 <sup>o</sup>	<sup>o</sup>	.761 <sup>o</sup>
	N <sup>o</sup>	9 <sup>o</sup>	9 <sup>o</sup>	9 <sup>o</sup>	9 <sup>o</sup>
d <sup>o</sup>	Pearson Correlation <sup>o</sup>	-.387 <sup>o</sup>	.298 <sup>o</sup>	-.119 <sup>o</sup>	1 <sup>o</sup>
	Sig. (2-tailed) <sup>o</sup>	.304 <sup>o</sup>	.436 <sup>o</sup>	.761 <sup>o</sup>	<sup>o</sup>
	N <sup>o</sup>	9 <sup>o</sup>	9 <sup>o</sup>	9 <sup>o</sup>	9 <sup>o</sup>

Figure 17: NSI Correlations with Climate Factors

Correlations <sup>o</sup>		MLI <sup>o</sup>	WFI <sup>o</sup>	NRI <sup>o</sup>	EEI <sup>o</sup>
t <sup>o</sup>	Pearson Correlation <sup>o</sup>	-.676* <sup>o</sup>	-.488 <sup>o</sup>	-.733* <sup>o</sup>	-.415 <sup>o</sup>
	Sig. (2-tailed) <sup>o</sup>	.046 <sup>o</sup>	.182 <sup>o</sup>	.025 <sup>o</sup>	.266 <sup>o</sup>
	N <sup>o</sup>	9 <sup>o</sup>	9 <sup>o</sup>	9 <sup>o</sup>	9 <sup>o</sup>
p <sup>o</sup>	Pearson Correlation <sup>o</sup>	.255 <sup>o</sup>	.372 <sup>o</sup>	.581 <sup>o</sup>	.393 <sup>o</sup>
	Sig. (2-tailed) <sup>o</sup>	.509 <sup>o</sup>	.324 <sup>o</sup>	.101 <sup>o</sup>	.296 <sup>o</sup>
	N <sup>o</sup>	9 <sup>o</sup>	9 <sup>o</sup>	9 <sup>o</sup>	9 <sup>o</sup>
d <sup>o</sup>	Pearson Correlation <sup>o</sup>	-.749* <sup>o</sup>	-.725* <sup>o</sup>	-.314 <sup>o</sup>	-.691* <sup>o</sup>
	Sig. (2-tailed) <sup>o</sup>	.020 <sup>o</sup>	.027 <sup>o</sup>	.410 <sup>o</sup>	.039 <sup>o</sup>
	N <sup>o</sup>	9 <sup>o</sup>	9 <sup>o</sup>	9 <sup>o</sup>	9 <sup>o</sup>

\*. Correlation is significant at the 0.05 level (2-tailed).<sup>o</sup>

\*\*. Correlation is significant at the 0.01 level (2-tailed).<sup>o</sup>

Figure 18: NSI Indicators Correlations with Climate Factors