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## **Abstract**

Rare earth minerals and metals will become a key commodity for the energy transition. Hence, benchmarking energy technologies based on the real costs and pollution associated with the mining, building, and recycling renewable energy technologies can give us the true implications the energy transition will have on our societies. The results from the energy model presented on this paper show renewable energy technologies a strong supporter of the energy supply but a weak stand-alone provider. Furthermore, high upfront costs and Global Warming Potential, GWP figures may have negative implications to the Human Development Indicators. Finally, the energy transition may induce elevated prices of rare earth minerals and metals, as these resources are heavily relied upon the production of renewable energy technologies.

Keywords: Energy system model, Global Warming Potential, Capacity factor, Life-cycle cost and GWP allocation, Rare-earth minerals, Human Development Index

## Content:

1. [Introduction](#)
2. [Literature on Energy models and author contribution.](#)
3. [Roadmap](#)
4. [Model](#)
  - 4.1 [Constraints](#)
  - 4.2 [Assumptions on constraints](#)
  - 4.3 [Variables](#)
  - 4.4 [Activity Distribution Map](#)
  - 4.5 [Calculating the maximum GWh production for each technology](#)
  - 4.6 [Solver](#)
5. [Life-cycle cost & GWP analysis and calculation](#)
  - 5.1 [Cost and GWP allocation](#)
  - 5.2 [Conversion of the cost and GWP from capacity to energy](#)
6. [Constraint Calculation Results](#)
  - 6.1 [Capacity Factor](#)
  - 6.2 [Cost; €M / GWh](#)
  - 6.3 [GWP; GWP/GWh](#)
7. [Cost and GWP minimization results](#)
  - 7.1 [Cost minimization scenario](#)
  - 7.2 [GWP minimization scenario](#)
8. [Energy consumption on the human development indicators](#)
9. [Raw Material Prices and Production implications](#)

**9.1** [Expected shock responses](#)

**9.2** [Metal & Mineral Data](#)

**9.3** [Metal Selection](#)

**9.4** [VAR diagnostics](#)

**9.5** [Causality Testing](#)

**9.6** [Impulse response function results](#)

**9.7** [Price Elasticity of Supply](#)

**9.8** [Alternative aggregate demand indicator](#)

**10.** [Conclusion](#)

Modeling the Energy Transition and analyzing its  
implications: A raw material approach

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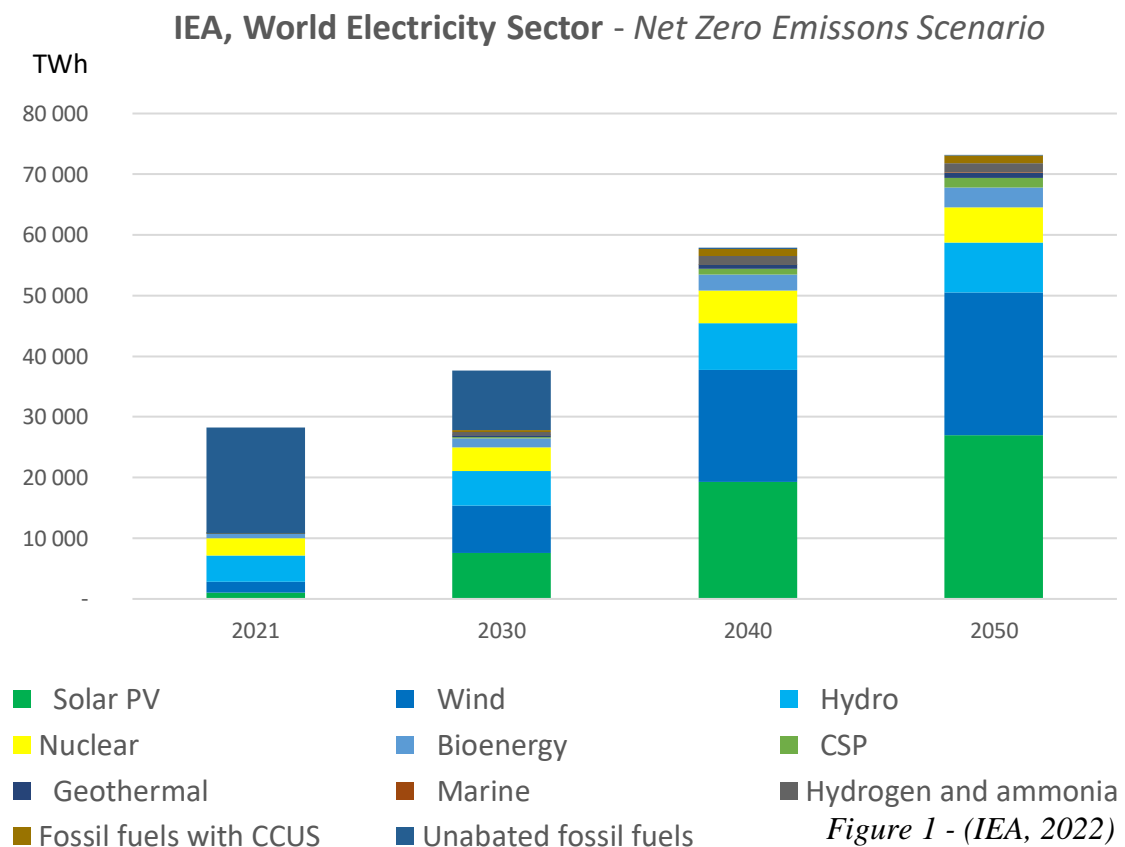
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# 1. Introduction

Amid the 2022 energy crisis caused by the war in Ukraine and climate change protests, governments are now investing in the energy transition in the hopes of higher energy security and a sustainable future. As of 2021, 44 countries and the European Union have pledged to net-zero-by-2050 policies. Altogether accounting for 70% of the total global emissions. –([IEA](#)) Out of all the countries ten of them have made the net-zero-by-2050 aspirations a legal obligation and another eight are in the process of doing so. In year 2021 Coal, Natural Gas, and Oil accounted for 62% of the total world energy production. – ([IEA](#)) Such fuels and technologies are cheap, abundant, and have low investment and maintenance costs, as we will discuss below. Therefore, they guarantee a stable economy and affordable energy prices.



The International Energy Agency net-zero-by-2050 report lays out a blueprint for each union member to follow and it serves as a reference for many others. What the report predicts is that by 2050 the world's 88% electricity needs will be met by renewable energy, mainly Solar PV, Windmills, and Hydro. Nuclear will account for 8% and the other 4% will be met by carbon capture coal and hydrogen. –([IEA](#))

Having seen the scale at which the IEA member countries and others are aspiring to expand renewables towards reaching the net-zero-by-2050 goals I raise the following questions:

1. How would the electric energy supply of a country look like constrained by the pollution and cost of each technology with emphasis to raw materials?
2. What are the implications of the transition on human development indicators and raw materials supply chains and prices?

In this paper, I propose building an energy system model for a country and map out how the energy supply would look like, based on cost, and pollution constraints. As a second part of the paper, I lay out the implications of the energy transition and analyze. The model requires a specific country as an example to increase the level of accuracy through official governmental data. The country of choice for this paper is Switzerland due to the data availability and the high-to-medium share of renewable energy resources like wind, sun, and precipitation.

## **2. Literature on Energy models and author contribution.**

Energy Models can be separated into two main categories: bottom-up and top-down approaches:

- A bottom-up model collects highly detailed technical data from energy technologies mentioning; capacity factors, production functions/patterns, or economic such as life-cycle investing and maintenance costs. The end-use demand is taken as given. A bottom-up approach is a mathematical formulation in which a function computes a set of variables that minimize or maximize an objective function subject to a set of constraints. This approach aims to model an energy system or sector to understand how technological changes or advancements affect the energy supply structure. –([Herbst](#))
- A top-down approach collects high-level information on different activities not necessarily related directly to energy technologies or consumption, such as economic activity, trade activity, and food prices, and associates that data with the corresponding energy consumption or structure. This approach aims to understand the implications of an energy structure on wider economic factors. –([Herbst](#))

My model of choice is a bottom-up approach which gives the ability to understand the true capacity of a country to adopt the net-zero-by-2050 aspirations and later to quantify the final energy delivered to the end-use consumer. The model is easy to understand and is explained in plain English. As part of my contribution, I will analyze the potential effects my model's results will have on the human development indicators and risks related to supply chains and prices of raw materials needed on building the technologies.



### 3. Roadmap

In the first section of the paper, I introduce the subject of my paper, points 1, 2. In the second section after the introduction and the author's contribution, I lay out how the model has been constructed and all the relevant assumptions. Section 2 includes points 4, 5, and 6. In section 3 which includes points 7, 8, 9, I discuss and lay out the implications of the energy transition on the human development indicators and on the rare-earth metals and minerals. Point 10 are the concluding remarks of the paper. All the data and excel models used on this paper is are open source and can be downloaded via the GitHub link on the reference section.

### 4. Model

Switzerland is the country with the lowest carbon footprint among the IEA members. Their electricity sector is dominated by nuclear 30% and hydro 60% generation as of 2021- (IEA). Being a country, which has its domestic renewable energy technologies resources exploited to a large extent, makes Switzerland a great case study as it minimizes the chances of introducing new technologies that were not included in this study. Energy technologies that will be included in the Switzerland case study are Oil, Natural Gas, Biofuels, Waste, Nuclear, Hydro, Solar PV, and Windmills.

#### 4.1 Constraints

- *f Min* represents the minimum amount in GW of a particular energy source that must be included in the model. For the net-zero-by-2050 scenario, I will assume the minimum is 0.
- *f Max* represents the maximum domestic capacity of a technology that can be deployed given in GW. In more concrete terms, the amount of hydropower

plants is constrained by the number of rivers and precipitations available within a country. – ([Bieler](#))

- **Cf**, capacity factor represents the ratio between the energy produced and the total capacity of a particular energy technology installed. In other terms, this ratio gives the effectiveness at which a primary energy source gets converted into usable end-use energy. It accounts for both the effectiveness of a particular technology and the abundance of a primary resource. We would expect a higher capacity factor for Solar PV in Dubai than in London due to higher solar radiation. But the city of London may have installed more efficient panels which can even out the difference. Therefore, each calculation must be performed for every country individually. Given by (1):

$$(1) Cf = \frac{GWh}{(GW \times 8766h)}$$

- *GWh represents the end-use energy produced in one year and GW represents the maximum capacity of production at a given hour, multiplied by the total hours of one year.*

- **Costs** are categorized into investment; **C<sub>inv</sub>** and maintenance costs **C<sub>main</sub>**. Being self-explanatory costs are expressed in millions of euros per GW capacity **M€ / GW**. – ([Muyldermans](#))

- **GWP** stands for “Global Warming Potential” which expresses the potential for greenhouse gas emissions to trap heat in the atmosphere. It measures how much energy is trapped in the atmosphere due to one unit of a certain gas being

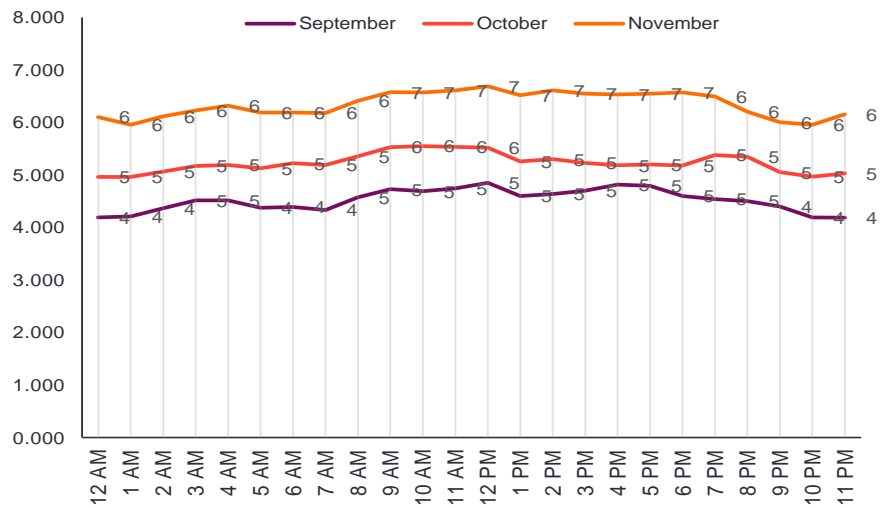
released compared to the emission of one unit of carbon dioxide. Expressed as megatons of Co<sub>2</sub>, *Mt-Co<sub>2</sub>*. – ([ScienceDirect](#))

## **4.2 Assumptions on constraints.**

The assumption of this paper has been kept at a minimum to increase the accuracy of the results. Nonetheless, they can be crucial when dealing with ambiguity. The maximum capacity that can be installed for nuclear, fossil fuels, waste, and biofuels has been assumed to be the current capacity as of 2021. This assumption is backed by the Swiss government's decision on phasing out nuclear and fossil fuels. – ([Swiss Confederation](#)) For waste I assume there is a saturation since this is dependent on limited raw materials (waste). Biofuels I assume that there is a saturation as well due to a large amount of land and other resources it takes to produce them. Such resources are also limited by the demand for agriculture. On the other hand, the minimum amount for each energy source is 0. The point of the model is to create a pathway towards future investments. Considering current installed capacities for each technology as minimum values will not produce any results.

## **4.3 Variables**

In this study, I have adopted a 12 typical days approach for a whole year based on the methodology of EnergyScope TD by –([Limpens](#)). Each typical day represents a month, and every hour of that day is the average consumption or production of every day of that month. This methodology gives breaks down the supply function on monthly and hourly intervals also simplifying the model's output complexity. The variables optimized subject to the constraints are every hour of a typical day. Figure 2 on the side represents the hourly consumption of three typical days of January, February, and March.



**Figure 2. Hourly GW consumption**

## 4.4 Activity Distribution Map

Activity maps for the demand side of the electricity sector capture how the demand changes from one hour to another and between months. It shows where the peaks are, lows, and everything in between for a whole year given in 12 typical days for each month. Activity maps for the demand side are straightforward to understand, however, that is not the case with the supply side of the electricity sector.

Energy technologies such as oil, gas, biofuels, waste, or nuclear plants are simple to model if we only are given the capacity of the plant. This is because the primary fuels to run these plants can be controlled by us. But in the case of renewable primary sources such as sun radiation, wind, and precipitations the data provided by the national weather stations can be ambiguous for the model to give precise solutions. I propose mapping the activity of each renewable energy source into an activity map to understand how annual production is distributed between a typical day of each month and hours.

The data needed to map solar PV and wind activity has been provided by the “Open-Source System Data” -([OPSD](#)). They have a wide range of time-series data for every country in the EU. The hourly time-series data starts in 2014 with a capacity of 455 GW and ends at the end of 2019 at 620 GW. Since we want to map only the activity of production then we must adjust production to one specific capacity to later get the average of each hour. Production of both wind and solar are then adjusted to their first year on the time series production data according to formula (2):

$$(2) \text{ Adjusted capacity} = \left(1 - \frac{(GW - GW_{\alpha})}{GW_{\alpha}}\right) \times GWh$$

- ***GW<sub>α</sub>*** represents the anchor value (the year 2014 capacity) for adjusting the rest of the production values.

After adjusting each time series to a single capacity production, we can then take the average production of each hour for every day of the month. Every hour of each typical day is then divided by the total GWh production giving the production distribution. Tables 2 & 3 give how the production changes for each hour of the typical day/month.

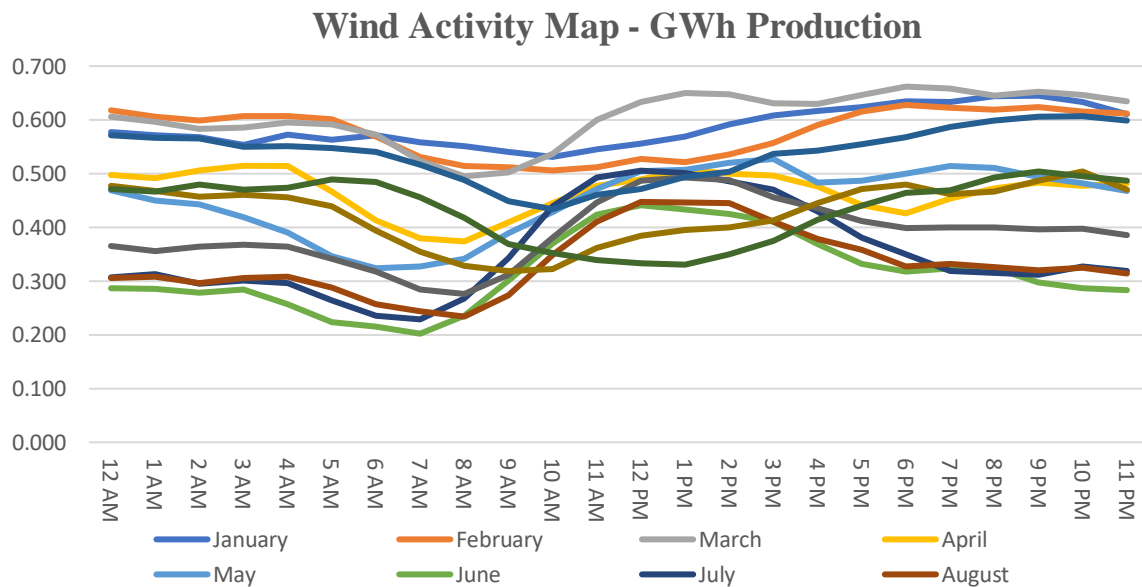


Figure 3 - (Open Power System, 2023)

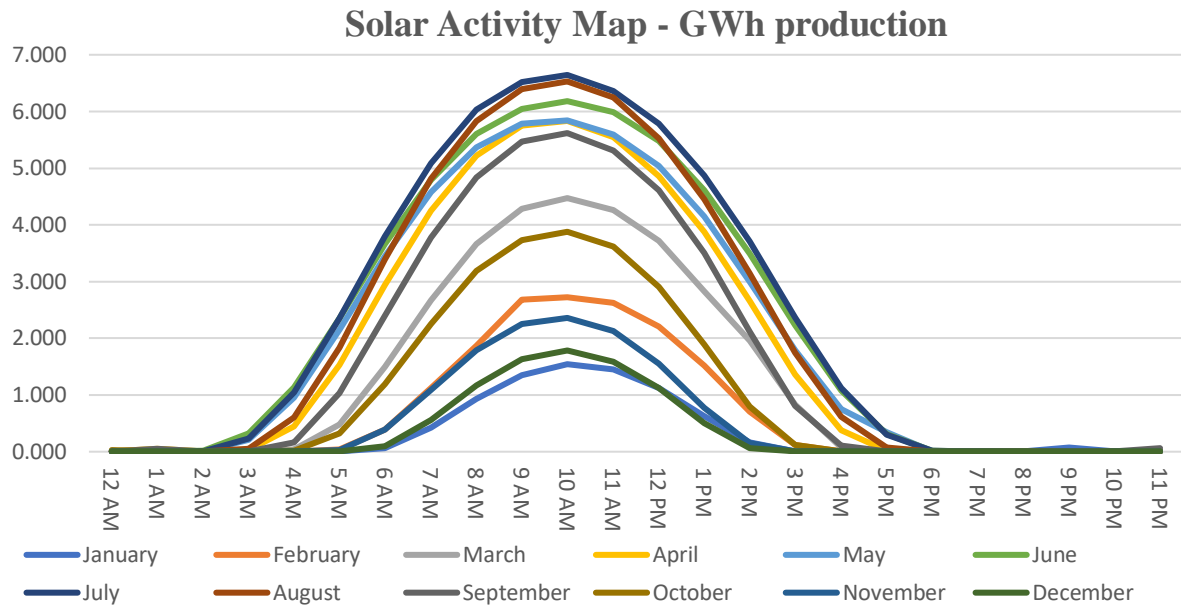
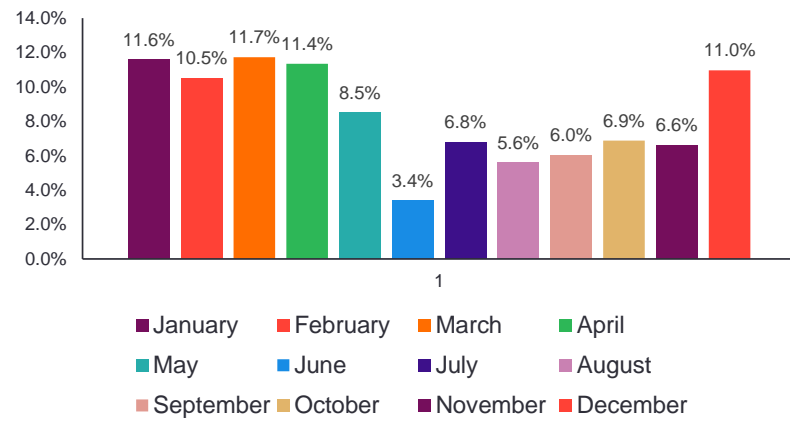


Figure 4 - (Open Power System, 2023)

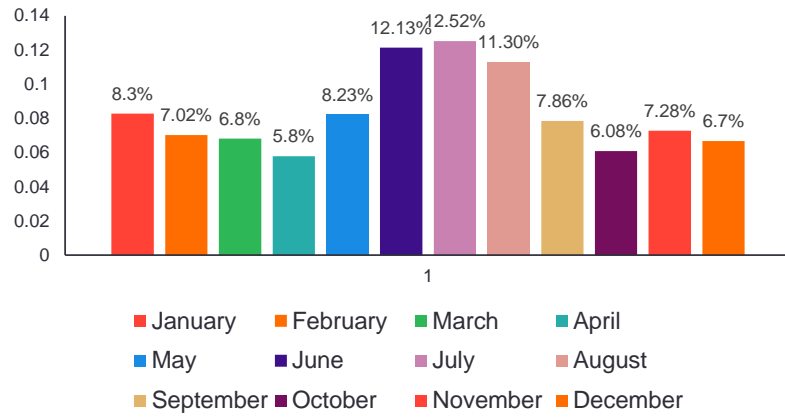
Monthly hydropower production data have been taken from official Swiss sources. – ([Swiss confederation](#)). The reason why in this study I consider hourly time-series for hydro production ambiguous is that dams serve as energy storage which can be deployed at any given time. In other words, if we get a large number of rainfall or other precipitation then the dam will use the collected water when it needs to during the month. This means that each typical day will have different values but each hour will have the same value within a day. For nuclear production, I use the same reasoning as hydro data.

**Nuclear Production Distribution**



*Figure 5 – (Open Power System, 2023)*

**Hydro Production Distribution**



*Figure 6 – (Open Power System, 2023)*

For Waste, Natural Gas, and Oil there was no monthly electricity production data available due to the small share they carry. Therefore, I have included them in the model as constants for every hour of each typical day. A distribution map gives each hour a percentage value of total GWh production for the 12 typical day intervals. Such a distribution map can show us how a certain amount of GWh is expected to be distributed across 12 typical days and every hour of each day for some technologies.

#### 4.5 Calculating the maximum GWh production for each technology

Although the maximum production data is provided by the Swiss sources in annual GWh production I have to convert it to the 12 typical day interval production we are considering. First I calculate the capacity that would be required to match the given production according to (3)

$$(3) \quad GW_{\alpha} = \frac{GWh/Cf}{8766h}$$

- 8766h represents the number of hours in one year.

After calculating the capacity of each energy source I convert this capacity into GWh production for the 12 typical days we have considered. Given by (4):

$$(4) \quad GWh = GW \times h \times Cf$$

- *To estimate the expected GWh production from a given available capacity, we have to multiply it by the running time of this capacity in total hours, and again multiply it with the capacity factor Cf, which accounts for the inefficiencies in either resources or and technological. In our model, I am considering only 12 typical days which account for 288 hours of running time. The capacity factor Cf for each technology is given in table 1 and it has been calculated from historical production data via formula (1).*

## 4.6 Solver

To solve the model I have used the MS Excel Solver. The Excel “Simplex LP” Solver is very easy to use and intuitive but very time-consuming. It can only calculate 200 variables at a time and my model consists of 2300 variables. To solve the model in Excel Solver I had to calculate each typical day separately. The values used for the calculation are summarized in table 1.



## 5. Life-cycle cost & GWP analysis and calculation

The GWP index and costs for each technology were given by the – ([Muyldermans](#)) in the form of GWP/GW and €/GW capacity. In other words, it estimated how much 1 Gigawatt of each technology costs and its Global Warming Potential (GWP) value. The cost was separated into investment and maintenance. Furthermore, the costs and GWP values were given as a total for each technology's entire life cycle. When doing the analysis and calculation of costs and GWP the two key determinants were:

- Lifecycle of the technology.
- The efficiency of the technology to convert raw resources into energy.

Both points above will be factored into the final estimated values on points **5.1** and **5.2**.

### 5.1 Cost and GWP allocation

Cost and GWP allocation are done on a life-cycle basis. Because we are only considering a 12 typical day interval, I propose allocating a portion of the cost and GWP index as a percentage value of the 12 typical day interval to its entire life cycle. Given by (5). Although the costs and GWP pollution are front heavy this solution can give us the real cost and GWP index because it accounts for the lifetime of each technology.

$$(5) \text{ Cost Allocation} = \frac{12TD}{365 \times X} \times \beta$$

- *12TD stands for 12 typical days*
- *X, represents the lifetime in years*
- *β, represents either total cost in €/GW or GWP/GW index*

Furthermore, another question may arise whether we consider the investment cost as a lump sum and allocate the 12TD interval only to the maintenance costs or whether take the sum allocate the 12TD interval to the total cost. I propose the latter, that is because we have a few technologies that have high investment costs but also a longer life cycle. By summing investing and maintenance costs to later allocate the 12TD interval I can get a real cost per gigawatt.

## **5.2 Conversion of the cost and GWP from capacity (GW) to energy (GWh)**

The cost, €/GW and GWP, GWP/GW expresses each value in terms of capacity installed in Gigawatts GW. Converting them into energy delivered such as €/GWh and GWP/GWh is very important, not only for estimating the hourly cost and pollution of the model but mainly to account for the efficiency of each technology to convert raw resources into energy. In concrete terms, one technology may have cheap costs but it's not efficient enough to convert raw resources such as wind, solar radiation, or precipitations into energy. The reasons can be either local or geographical.

I propose calculating the total cost of the installed capacity and then dividing that total cost by the amount of GWh produced during the 12 typical day interval. Illustrated into (6):

$$(6) \quad \beta / GWh = \frac{GW * \beta}{GWh}$$

- *C represents the allocated cost during the 12TD interval.*
- *GWh represents the total amount of energy delivered for end-use.*
- *$\beta$  represents either allocated cost, €/GW, or GWP index, GWP/GW.*

Furthermore, we know from (4) that  $GWh = GW \times h \times Cf$  hence:

$$\beta / GWh = \frac{GW * \beta}{GW * h * Cf}$$
 After doing the algebraic simplification of the formula above I

arrive at:

$$(7) \quad \beta / GWh = \frac{\beta}{Cf * h}$$

- In our case, the 12 typical days account for 288 hours, hence,  $h = 288$ .

Number (7) shows us that both values €/M/GWh and GWP/GWh are dependent on constants. Such calculation can capture the efficiency of an energy technology via the Capacity Factor Cf. We can now say that our final values of €/M/GWh and GWP/GWh account for the lifecycle and the efficiency of each technology.

Table 1.

Electricity Sector	f min [GW]		f max [GW]		Costs				GWP/GWh	GWP Allocation	€/MWh	Cf capacity Factor	GWP ktCO2/GW	Lifetime	Current Installed Capacity [GW]
		€	€	-	€	C inv M€/GW	C main M€/GW	Total							
Coal	0	0	€	-	€	-	€	€ -	0.400	-	€ -	0	-	0	0
Oil	0	18	€	55	€		€	€ 56	0.315	-	0.0007	0.513	-	17	0.004
Natural Gas	0	535	€	1,408	€		€	€ 1,499	0.260	-	0.0231	0.211	-	35	0.289
Biofuels	0	693	€	1,374	€		€	€ 1,486	0.033	-	0.0375	0.452	-	10	0.1748
Waste	0	2081	€	2,928	€		€	€ 3,036	0.150	-	0.0204	0.585	-	29	0.406
Nuclear	0	19364	€	4,846	€		€	€ 4,948	0.002	0.3671	0.0128	0.736	670	60	3
Hydro	0	49395	€	5,000	€		€	€ 5,050	0.027	2.2533	0.0484	0.291	2810	41	15.66
Solar PV	0	12000	€	870	€		€	€ 880	0.054	1.4584	0.0427	0.094	1109	25	3.482
Wind	0	4000	€	1,040	€		€	€ 1,043	0.017	0.9156	0.0316	0.188	557	20	0.0884

## 6. Constraint Calculations

The inputs that I have calculated are, *Capacity Factor Cf*, *GWP/GWh*, and *€/GWh*. In the following paragraphs, I will do a summarization of these results.

### 6.1 Capacity Factor

In terms of efficiency in converting raw resources such as solar radiation, uranium, or oil into energy. On average fossil fuels have the higher scores due to their energy-dense structure. Nonetheless, my results give nuclear power plants the highest score of 73.6%, highest score from fossil fuels by a large margin. This explains the high energy density of Uranium as a material. Second comes waste followed by oil, biofuels, and natural gas at 58.5%, 51.3%, 45.2%, and 21.1% respectively. Renewable energy scores are led by hydropower plants at 29.0%, wind at 18.8%, and solar at 9.4%. These values not only they describe the efficiency of the technologies we are considering but also, they capture the geographical location of where these technologies are installed. For solar we would expect to see a higher score in a country with more solar radiation.

### 6.2 Cost; €/M / GWh

In terms of cost, the GWh fossil fuels lead the way with Oil at €734 per GWh. The second cheapest comes Nuclear at €12,785. That is followed by Waste at €20,439, Natural Gas at €23,144, Wind at €31,592, Biofuels at €37,508, Solar at €42,718, and the most expensive Hydro at €48,381. The cost per GW capacity of Hydro and Nuclear are quite similar at €M 5,050 and €M 4,948 per GW capacity but the reason why my methodology makes hydro a

lot more expensive is the short lifecycle of 41 years compared to 60 for Nuclear and lower capacity factor as discussed on point 5.1.

### 6.3 GWP; GWP/GWh

GWP scores will be given in MWh to make them easier to grasp. Nuclear is the least pollutant with a score of 1.7/MWh. After that comes Wind, Hydro, Biofuels, and Solar with scores of 16.9/MWh, 26.9/MWh, 33.0/MWh, and 53.8/MWh, respectively. Then, fossil fuel scores are in the hundreds of units. Waste, Natural gas, Oil, and Coal have their scores at 150/MWh, 260/MWh, 315/MWh, and 400/MWh, respectively.

## 7. Cost and GWP minimization scenario results

Tables 2 and 3 summarize the results of the two scenarios. In the first scenario, we minimize Cost and in the second scenario, we minimize GWP.

### 7.1 Cost minimization scenario

In the cost minimization scenario, we see nuclear capacity at full utilization

*Table 2*

GWP Min	GWh	%GWh	Utilization	GWP	Cost per GW
Biofuels	-	0.0%	0.0%	-	-
Wind	131.4	9.1%	100.0%	2.2	4.2
Solar	-	0.0%	0.0%	-	-
Hydro	680.9	47.0%	42.0%	18.3	32.9
Nuclear	636.2	43.9%	100.0%	1.1	8.1
Waste	-	0.0%	0.0%	-	-
Natural Gas	-	0.0%	0.0%	-	-
Oil	-	0.0%	0.0%	-	-
<b>Total:</b>	<b>1448.5</b>			<b>21.6</b>	<b>€ 45.23</b>

accounting for 43.9% of the electricity supply. Hydro accounts for 25.9% of the supply followed by Solar and Wind at 13.5% and 9.1% respectively. Then, at a smaller percentage

come Waste, Biofuels, and Natural Gas at 4.7%, 1.6%, and 1.2%, respectively. All the energy

sources are utilized because Hydro being the most expensive has also a large capacity. Therefore, to minimize the cost you must substitute Hydropower with alternative sources such as fossil fuels which have smaller capacities of production. Hence, we see a diverse energy supply. The GWP for this scenario sits at 39.7 and the total cost of running the system generated is €41.5 million.

## 7.2 GWP minimization scenario

*Table 3*

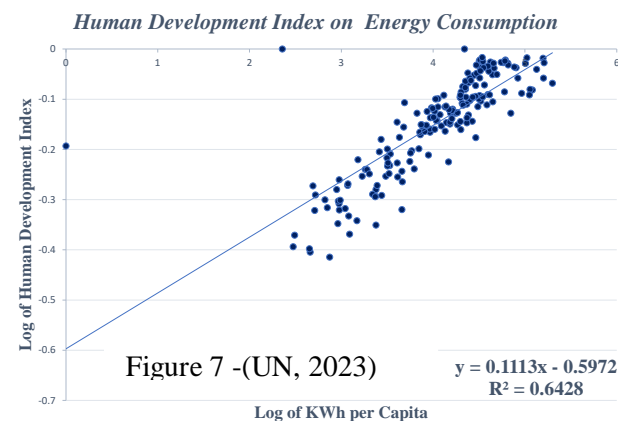
Cost Min	GWh	%GWh	Utilization	GWP/GWh	€/GWh
Oil	0.6	0.0%	100.0%	0.2	0.0
Solar	195.9	13.5%	49.7%	10.5	8.4
Wind	131.4	9.1%	100.0%	2.2	4.2
Biofuels	22.8	1.6%	100.0%	0.8	0.9
Natural Gas	17.6	1.2%	100.0%	4.6	0.4
Waste	68.4	4.7%	100.0%	10.3	1.4
Nuclear	636.2	43.9%	100.0%	1.1	8.1
Hydro	375.7	25.9%	23.2%	10.1	18.2
<b>Total:</b>	<b>1,448.5</b>			<b>39.7</b>	<b>€ 41.5</b>

On the Global Warming Potential index minimization, we notice fewer energy sources being utilized. This is because Hydro and Nuclear are capable of large capacities accounting for 43.9% and 47% of the energy supply. The wind is the third and last source accounting for 9.1% of the energy supply. Utilization for Nuclear, Hydro, and Wind is at 100%, 42%, and 100% respectively. The GWP index for this scenario sits at 21.6 which is half of the cost min scenario and the cost of running the system is slightly higher at €45.2 million. The reason why we see a smaller difference in costs between the scenarios is that the cheapest technologies available have very limited capacity. Point out, Nuclear is a key player in both cost and GWP but the Swiss government's decision to phase out nuclear and

other non-renewable technologies will further increase costs. The costs that are included in this model constitute the construction and maintenance costs which are the investment we must make to extract the energy we are aiming for. Higher costs for the same amount of energy delivered means we will have less net energy available to us. The relationship between energy delivered and energy invested is captured by the Energy Return on Investment (EROI) ratio. The literature on this metric is vast.

## 8. Energy consumption on the human development indicators

On a published paper titled “Energy, EROI, and quality of life” the author, [Lambert](#) does an extensive research and analysis on how energy consumption correlates with quality-of-life indicators. Given the rising costs of drilling for oil the author analyses the EROI at a societal level and energy-per-capita and finally



composes a new energy index. The author also explains how the price of a fuel is determined by its embodied energy or by the energy required to bring the raw material to its final consumption stage. Their new index “LEI” which combines, the energy intensity of a GDP unit, energy consumption per capita, and Gini-index is found to be correlated with quality-of-life indicators. I want to take this research further by testing for causality of energy consumption on human development indicators.

The human development index (HDI) was developed and calculated by the “United Nations” as part of its development program. It was first introduced in 1990, and its goal is to emphasize people and their capacity only and not include the economic indicators alone.



– ([UNDP](#)) The human development office computes the index for about 190 countries, and it uses the following indicators:

1. Life expectancy at birth
2. Expected years of schooling
3. Mean years of schooling
4. Gross national income per capita (PPP \$)

The human development index is the geometric mean of its indicators which alone are highly correlated with the energy consumption levels for the sample of 182 countries I have included in my study. On the right, I have presented the graphs and the regression line equations of energy per capita and the HDI index followed by its composite indicators. The energy per capita is given in KWh and it includes every energy sector such as electricity, industry, transportation, and more, provided by "Our World in Data" – ([Our World in Data](#))

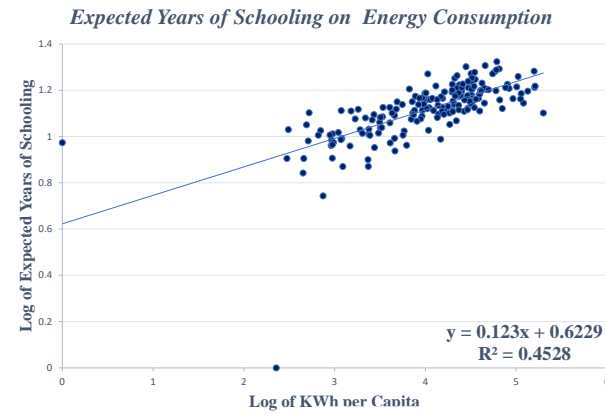


Figure 8 -(UN, 2023)

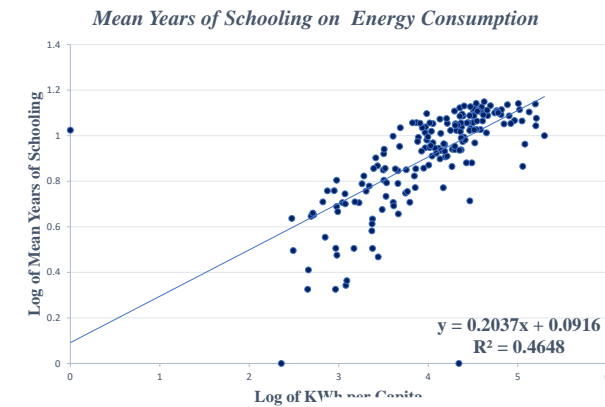


Figure 9 -(UN, 2023)

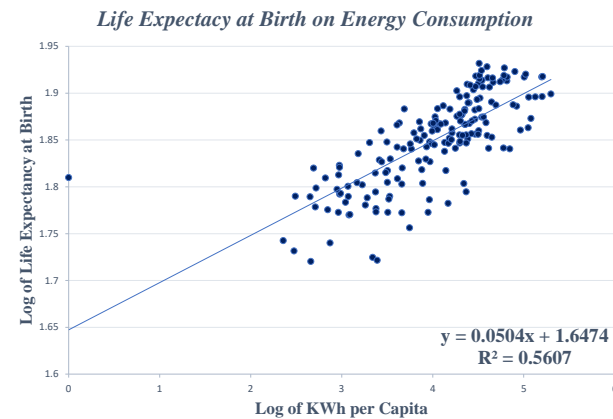


Figure 10 -(UN, 2023)

It is common to think that the correlation is biased due to the economic development affecting positively the HDI indicators and energy consumption. But my hypothesis suggests that energy consumption precedes economic development and ultimately human development. To test this hypothesis, I propose performing an instrumental variable analysis on the Human Development Index as the dependent variable

and testing for endogeneity on the energy consumption variable. I cannot perform an endogeneity test on energy consumption with the index itself as dependent on its statistical stationery. The only values the index can take are between 0 and 1. Hence I will only perform tests on its indicators.

The instrumental variable that I propose is the difference between the average minimum and maximum temperature for each country, calling it temperature variation. The data minimum and maximum average temperature were provided by the World Bank at their “Climate Change Knowledge” database portal. – ([World Bank](#))

The reasoning behind this instrumental variable is that countries with a large difference in the temperature range consume more energy than other counties with a smaller range, no matter the actual temperature values. Explained in another way, such countries would be the ones with extreme winters and summers, and their inhabitants would require a large amount of energy for both heating in the winter and cooling in the summer. The temperature range proves to be significant at the 95% confidence interval with a coefficient of 0.75 and an r-squared value of 18.5%. A such instrumental variable is correlated with the outcome/dependent but does not have a causal effect on the outcome due to its randomness.

To further strengthen my model, I choose to control for some economic development indicators that drive both energy consumption and human development. Such control indicators are GDP/capita and gross national income. GDP/capita is an economic indicator driven by energy consumption and has a positive effect on human development indicators. Gross national income accounts for wealthy countries that can afford to consume large amounts of energy and invest in improving human development indicators. In my hypothesis energy demand is infinite and it's only constrained by supply, hence

demand/consumption equals supply. For running the instrumental variable regression, I have used the Two-stage least squares (2SLS) method. The structural equation:

After using the instrument to predict the energy consumption KWh/Capita the equation is as (8):

$$(8) \log(HDI\ Indicators) = \beta_0 + \beta_1 \log \frac{K\hat{W}h}{Capita} + \beta_k X_k + \varepsilon$$

- $\frac{K\hat{W}h}{Capita}$  represents the predicted energy consumption based on the “Temperature\_Variation” instrument.
- $k$  stands for: National Income, GDP per capita

After running the regression and testing for endogeneity, results show evidence of exogeneity of the KWh/Capita variable for all three indicators tested. Below you will find a summary table of the results.

Table 4

Indicators	Durbin Score	Wu-Hausman Score	Min Eigenvalue	Coef.
Life Expectancy at birth	0.72	0.73	8.73	0.034
Expected Years of Schooling	0.87	0.88	8.46	0.053
Mean Years of Schooling	0.77	0.77	8.46	0.149

Furthermore, by testing for weak instruments, the minimum eigenvalue showed above we can reject the null hypothesis that our instruments are weak at the 90% confidence interval. The strongest coefficient is the one between energy consumption and mean\_years\_of\_schooling followed second by expected years\_of\_schooling and life\_expectancy\_at\_birth.

## 9. Raw Material Prices and Production implications

Green energy technologies, such as Solar, and off/on-shore wind power, can improve energy security by capitalizing on a country's natural resources. However green energy technologies are highly dependent on the raw materials needed to build them. The net weight of raw materials per MW of green tech is substantially higher. Moreover, a large portion of these materials consists of rare earth metals and minerals. The figure 11, below shows a contrast between the nonrenewable and the green technologies in terms of amounts needed and type of materials.

### Raw Materials, Kg/MW

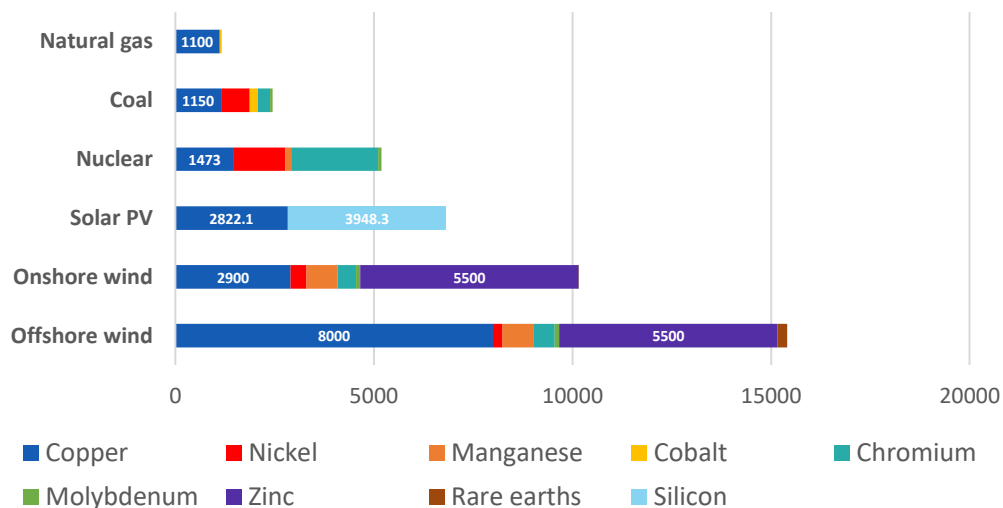


Figure 11 – (IEA, Critical Minerals, 2021)

Referring to figure 1, the international energy agency (IEA) has estimated that by 2050, around 70% of the total world's electricity will be produced by wind, solar, and hydro. These goals demand a substantial number of raw materials and rare earth minerals to build the technologies and infrastructure, far beyond the current production levels.

A transition may induce a shock in material prices because of a shortfall in supply. The supply chains of the raw materials will also be a crucial factor in the energy transition, particularly after we witnessed the 2022 energy crisis, triggered by a disruption in a heavily reliant Russian supply chain. Such disruption caused elevated energy prices and high inflation. In the case of raw materials for the energy transition, China holds a more dominant position than Russia did with natural gas. Another aspect as important as the others are the role that these materials play in non-energy-related industries.

As such, I lay out three main raw material risks that the energy transition will encounter. Those are:

1. Positive Demand shock, caused by rapid investing in green technologies
2. Positive non-energy industrial shock, caused by an expanding business cycle
3. Positive raw materials supply shock, caused by countries investing in production and new supply chains.

Demand shocks can be captured by the price of the raw materials. Non-energy industrial shocks can be captured by industrial production indexes, more on the index chosen below. Lastly, a supply shock is captured by the production figures of the respective raw material. To understand the implications of these shocks I propose constructing a Vector Autoregressive model with price, production of the respective metal, and industrial production as endogenous variables. The three shocks I laid out above will be treated as exogenous.

## 9.1 Expected shock responses

*Table 5*

	Price of Raw Material	Production of Raw Materials	Industrial Production
Demand Shocks	+	+	0
Supply Shock	-	+	0
Industrial Shocks	+	+	+

## 9.2 Metal & Mineral Data

The data for the metal price and production has been provided by the United States Geological Survey ([USGS](#)). For most of the metals, the time series annual data for price and production can go back to 1900 and as recently as 2019. Industrial production has been provided by the Federal reserve bank of ST. Louis statistics ([FRED](#)). It measures the real output of all relevant establishments located in the United States, regardless of their ownership.

## 9.3 Metal Selection

For the metal selection, I propose building a risk score index consisting of supply chain risk, end-of-life recycling score, and the percent demand growth of the respective metal from current production levels. The total risk score is calculated as the geometric mean of the three indicators. More on the indicators:

1. The supply chain risk score is the weighted average of each country's government stability index based on the contribution they have to the total supply of the respective metal. Data for the supply chains are provided by the United States Geological Survey (USGS). The government risk score values vary between 0-100 and have been provided by World Bank. ([World Bank](#))
2. End of Life (EoL) recycling rates were taken from the United Nations Environment Program Report ([UNEP](#)). End of Life recycle rates describe the amount that is recycled/retained. I have converted the rate as a risk score by capturing the amount lost that is;  $1 - \text{EoL}\%$ .

3. The % growth from current levels of production has been provided by the World Bank in its "[Minerals for Climate Action](#)" report.

Below you will find the table for each metal included in the study case: A key fact to point out is the high supply chain risk score each metal has. This is due to the large quantity of each metal that China is supplying being a country rich in resources and highly unstable politics.

*Table 6*

Metal:	Supply Chain Risk:	Recycling Risk Score:	% Growth Amount Requirements:	Risk Score:
Lithium	85.77	99	488	2.18086
Indium	67.06	99	231	2.05001
Graphite	37.74	100	494	2.04043
Vanadium	66.21	99	189	2.02244
Cobalt	85.70	32	460	1.97870
Zinc	83.26	58.5	98	1.89059
Nickel	86.72	41	99	1.83948
Silver	75.86	41	56	1.74453
Neodymium	47.79	99	37	1.73862
Lead	87.84	28	18	1.52455
Molybdenum	53.76	70	11	1.49257
Titanium	53.71	9	64	1.43934
Aluminium	51.50	43	9	1.38604
Copper	65.97	52	7	1.38180
Manganese	88.42	47	4	1.25137
Iron Ore	69.84	28.25	2	0.93044
Chromium	47.04	10	2	0.79552

The chosen metals for the study are Vanadium, Indium, Zinc, and Rare Earth. The selection has been made based on the highest risk scores, data availability, and model stability. Rare Earth has not been included in the risk score table due to the lack of indicator data but knowing the importance of Rare Earth minerals I have decided to include it in the VAR model.

## **9.4 VAR diagnostics**

The diagnostics of the stability of the VAR model include eigenvalue stability condition and testing of autocorrelation via the langrage-multiplier test. The results for all VAR models are stable with all eigenvalues lying within the unit circle. Lastly, there is no evidence of autocorrelation at the selected lag.

## **9.5 Causality Testing**

To further understand the relationship of the shock I propose testing for causality via the Granger causality test. The results show that all metals and minerals included are endogenous and do not show a causal relationship with one-another. Only Vanadium results, show evidence of the causality of Price on Production. Such results of causality testing show that price, production, and economic activity are driven by other factors that are not included in the model. An exogenous shock such as a demand and production shock driven by changes in policies may cause economic instability due to unpredictable price and production volatility.



## 9.6 Impulse response function results

Overall, in terms of response dynamics for each shock, each VAR model is similar.

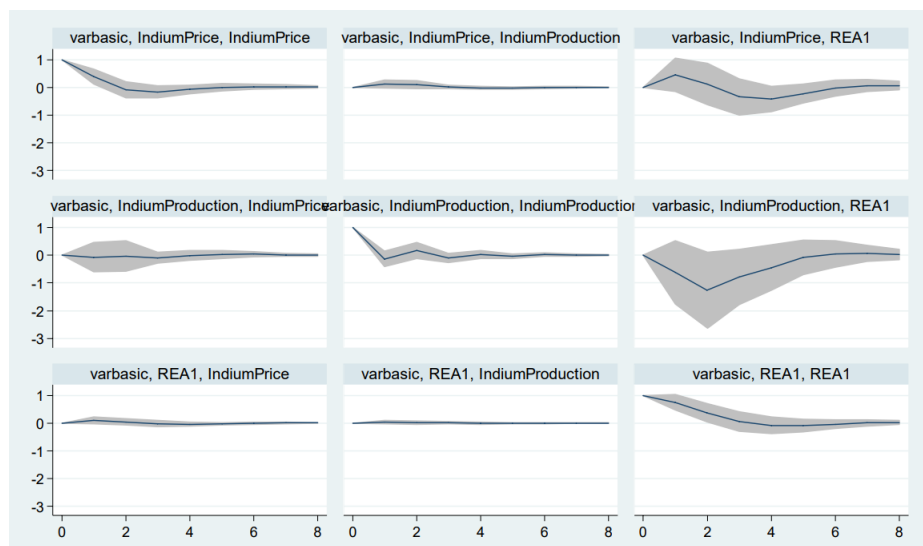


Figure 12 – Impulse Response Functions, Indium

Considering that each of our shocks is exogenous, our response function is solely induced by the shock we are testing for. A key result to point out is the low-price elasticity of supply when testing for a demand shock on production. The consistency of the supply elasticity results is maintained when examining the effect of a positive shock in aggregate demand on price and production. The results show a price level increase that surpasses the increase in production levels.

## 9.7 Price Elasticity of Supply

The table below summarizes the responses of price and production to a positive Price/Demand shock. Note that price and production responses are separate functions from one another and that they describe how would price and production respond to a one-unit positive shock in price.

Table 7.

Period:	Price Shock on Price (P) & Production (Q)							
	IndiumPP	IndiumPQ	RarePP	RarePQ	VanadiumPP	VanadiumPQ	ZincPP	ZincPQ
1	100.0%	5.1%	100.0%	1.6%	100.0%	11.5%	100.0%	0.9%
2	96.5%	6.3%	93.4%	2.0%	88.4%	11.7%	94.7%	4.1%
3	95.9%	9.9%	89.8%	1.6%	86.6%	11.8%	94.6%	4.3%
4	95.6%	9.8%	84.6%	1.5%	83.9%	11.8%	94.5%	4.5%
5	95.1%	9.8%	83.9%	2.3%	83.7%	11.8%	94.4%	4.5%
6	94.9%	9.9%	83.7%	2.9%	83.5%	11.9%	94.4%	4.5%
7	94.9%	9.9%	83.7%	2.9%	83.3%	11.9%	94.4%	4.5%
8	94.9%	9.9%	83.6%	2.9%	83.3%	11.9%	94.4%	4.5%

The responses of price on price show a slight decrease from the initial shock in period 1. As per production, it raises only by a small portion to the one unit increase in price. The response of production on a shock on price estimates for us the price elasticity of supply. These estimations are low and are also reflected in the elevated prices in all periods.

## 9.8 Alternative aggregate demand indicator

Using an alternative aggregate demand indicator such as Industrial Production we obtain similar results and a more obvious lag in production response is introduced. Between the two different aggregate demand indicators, the results stay consistent except for zinc which increases the amount produced.

Table 8.

Period:	Price Shock on Price (P) & Production (Q)							
	IndiumPP	IndiumPQ	VanadiumPP	VanadiumPQ	RarePP	RarePQ	ZincPP	ZincPQ
1	100.0%	5.8%	100.0%	0.0%	100.0%	0.2%	100.0%	11.7%
2	98.3%	7.1%	99.7%	0.0%	99.9%	1.6%	98.9%	12.7%
3	90.9%	10.4%	99.0%	15.8%	95.4%	1.5%	98.9%	13.2%
4	90.2%	10.4%	98.5%	16.1%	93.8%	3.8%	97.8%	14.1%
5	89.7%	10.3%	98.3%	16.0%	93.7%	4.5%	97.7%	14.2%
6	89.3%	10.4%	98.3%	16.0%	92.3%	4.7%	97.6%	14.3%
7	89.3%	10.4%	98.3%	16.2%	92.0%	5.1%	97.6%	14.3%
8	89.3%	10.4%	98.3%	16.2%	92.0%	5.4%	97.6%	14.3%

## **10.Conclusion**

Rare earth minerals and metals will become a key commodity for the energy transition. Hence, benchmarking these technologies based on the costs and pollution associated with the mining, building, and recycling renewable energy technologies can give us the true implications the energy transition will have on our societies.

Based on the model results, the renewable energy technologies show to be a strong energy provider for Switzerland in the long term. However, the high upfront costs per GW capacity, and the supply functions/patterns of renewable energy will require strong strategic investments towards rare-earth minerals supply-chains, and an overall update of the electricity grid. Furthermore, the Global Warming Potential (GWP) figures show that renewable energy pollution are front heavy, and we may risk reallocating the mining, building, and recycling pollution to other countries. Other risks as lower energy return on investment may have negative implication on the human development factors. Finally, elevated prices and supply shortfalls of rare earth minerals and metals show to be bottlenecks to the energy transition based on the IEA 2040 scenario.

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# Appendix

## Activity Distribution Map 1

WIND Activity Distribution MAP Switzerland													
Time	January	February	March	April	May	June	July	August	September	October	November	December	
12 AM	0.439%	0.471%	0.461%		0.379%	0.357%	0.218%	0.233%	0.232%	0.278%	0.363%	0.435%	0.358%
1 AM	0.434%	0.461%	0.453%		0.374%	0.342%	0.218%	0.238%	0.234%	0.270%	0.356%	0.431%	0.355%
2 AM	0.432%	0.455%	0.444%		0.385%	0.337%	0.212%	0.225%	0.225%	0.277%	0.348%	0.430%	0.365%
3 AM	0.421%	0.462%	0.446%		0.392%	0.319%	0.216%	0.229%	0.233%	0.280%	0.351%	0.419%	0.358%
4 AM	0.435%	0.462%	0.453%		0.391%	0.297%	0.196%	0.225%	0.235%	0.277%	0.347%	0.419%	0.361%
5 AM	0.428%	0.457%	0.450%		0.355%	0.264%	0.170%	0.201%	0.220%	0.260%	0.334%	0.416%	0.372%
6 AM	0.435%	0.433%	0.435%		0.314%	0.246%	0.164%	0.179%	0.196%	0.242%	0.300%	0.411%	0.369%
7 AM	0.425%	0.404%	0.399%		0.289%	0.249%	0.154%	0.174%	0.186%	0.217%	0.270%	0.393%	0.346%
8 AM	0.419%	0.391%	0.377%		0.284%	0.260%	0.179%	0.204%	0.178%	0.210%	0.250%	0.371%	0.318%
9 AM	0.411%	0.389%	0.382%		0.312%	0.296%	0.229%	0.261%	0.208%	0.237%	0.243%	0.341%	0.281%
10 AM	0.404%	0.385%	0.409%		0.339%	0.326%	0.282%	0.334%	0.265%	0.290%	0.246%	0.330%	0.268%
11 AM	0.414%	0.390%	0.457%		0.363%	0.358%	0.322%	0.375%	0.312%	0.340%	0.275%	0.351%	0.258%
12 PM	0.423%	0.401%	0.482%		0.372%	0.384%	0.336%	0.385%	0.340%	0.371%	0.293%	0.359%	0.253%
1 PM	0.433%	0.397%	0.495%		0.386%	0.385%	0.330%	0.381%	0.340%	0.375%	0.301%	0.375%	0.252%
2 PM	0.450%	0.408%	0.493%		0.381%	0.396%	0.323%	0.370%	0.339%	0.372%	0.304%	0.383%	0.266%
3 PM	0.463%	0.424%	0.480%		0.378%	0.401%	0.312%	0.358%	0.312%	0.347%	0.314%	0.409%	0.285%
4 PM	0.470%	0.449%	0.479%		0.362%	0.368%	0.281%	0.327%	0.288%	0.331%	0.339%	0.413%	0.316%
5 PM	0.475%	0.469%	0.492%		0.336%	0.370%	0.252%	0.290%	0.273%	0.313%	0.359%	0.422%	0.335%
6 PM	0.482%	0.478%	0.504%		0.324%	0.381%	0.242%	0.266%	0.249%	0.304%	0.365%	0.432%	0.353%
7 PM	0.482%	0.473%	0.501%		0.345%	0.391%	0.246%	0.243%	0.253%	0.304%	0.351%	0.446%	0.357%
8 PM	0.490%	0.471%	0.491%		0.360%	0.388%	0.247%	0.240%	0.248%	0.304%	0.355%	0.455%	0.375%
9 PM	0.491%	0.474%	0.497%		0.368%	0.376%	0.226%	0.237%	0.243%	0.301%	0.371%	0.461%	0.383%
10 PM	0.482%	0.469%	0.492%		0.364%	0.367%	0.218%	0.249%	0.247%	0.302%	0.383%	0.462%	0.376%
11 PM	0.464%	0.466%	0.483%		0.367%	0.356%	0.216%	0.242%	0.239%	0.293%	0.357%	0.456%	0.370%

## Activity Distribution Map 2

SOLAR Activity Distribution MAP Switzerland													
Time	January	February	March	April	May	June	July	August	September	October	November	December	
12 AM	0.0007%	0.0005%	0.0008%		0.0007%	0.0007%	0.0008%	0.0011%	0.0010%	0.0008%	0.0066%	0.0009%	0.0010%
1 AM	0.0007%	0.0005%	0.0008%		0.0144%	0.0007%	0.0008%	0.0011%	0.0009%	0.0116%	0.0009%	0.0009%	0.0010%
2 AM	0.0006%	0.0005%	0.0009%		0.0007%	0.0016%	0.0041%	0.0019%	0.0009%	0.0008%	0.0009%	0.0009%	0.0010%
3 AM	0.0007%	0.0005%	0.0009%		0.0065%	0.0525%	0.0814%	0.0575%	0.0130%	0.0009%	0.0008%	0.0008%	0.0010%
4 AM	0.0006%	0.0005%	0.0100%		0.1135%	0.2432%	0.2874%	0.2645%	0.1531%	0.0414%	0.0031%	0.0008%	0.0010%
5 AM	0.0007%	0.0090%	0.1219%		0.3871%	0.5454%	0.5963%	0.5959%	0.4660%	0.2615%	0.0812%	0.0073%	0.0010%
6 AM	0.0170%	0.0994%	0.3808%		0.7458%	0.8742%	0.9295%	0.9627%	0.8618%	0.6083%	0.3027%	0.0983%	0.0245%
7 AM	0.1064%	0.2845%	0.6773%		1.0774%	1.1605%	1.2153%	1.2885%	1.2204%	0.9566%	0.5720%	0.2750%	0.1424%
8 AM	0.2352%	0.4741%	0.9294%		1.3241%	1.3603%	1.4212%	1.5289%	1.4787%	1.2272%	0.8085%	0.4525%	0.2958%
9 AM	0.3431%	0.6801%	1.0856%		1.4575%	1.4672%	1.5328%	1.6531%	1.6210%	1.3867%	0.9449%	0.5709%	0.4126%
10 AM	0.3915%	0.6909%	1.1344%		1.4795%	1.4820%	1.5681%	1.6852%	1.6562%	1.4251%	0.9838%	0.5985%	0.4528%
11 AM	0.3675%	0.6670%	1.0800%		1.4068%	1.4175%	1.5187%	1.6133%	1.5843%	1.3476%	0.9167%	0.5393%	0.4027%
12 PM	0.2837%	0.5592%	0.9423%		1.2335%	1.2770%	1.3894%	1.4664%	1.4027%	1.1696%	0.7384%	0.3929%	0.2841%
1 PM	0.1592%	0.3849%	0.7166%		0.9848%	1.0512%	1.1690%	1.2363%	1.1302%	0.8886%	0.4793%	0.1977%	0.1288%
2 PM	0.0412%	0.1796%	0.4964%		0.6735%	0.7616%	0.8872%	0.9396%	0.7977%	0.5368%	0.2025%	0.0392%	0.0172%
3 PM	0.0016%	0.0315%	0.2118%		0.3439%	0.4526%	0.5665%	0.5996%	0.4436%	0.2044%	0.0301%	0.0015%	0.0011%
4 PM	0.0008%	0.0007%	0.0174%		0.0950%	0.1902%	0.2737%	0.2856%	0.1560%	0.0263%	0.0012%	0.0011%	0.0011%
5 PM	0.0007%	0.0006%	0.0008%		0.0045%	0.0877%	0.0800%	0.0770%	0.0178%	0.0008%	0.0009%	0.0010%	0.0011%
6 PM	0.0007%	0.0005%	0.0007%		0.0005%	0.0008%	0.0042%	0.0034%	0.0008%	0.0007%	0.0009%	0.0009%	0.0010%
7 PM	0.0007%	0.0005%	0.0007%		0.0005%	0.0005%	0.0006%	0.0009%	0.0008%	0.0007%	0.0008%	0.0008%	0.0010%
8 PM	0.0007%	0.0005%	0.0007%		0.0006%	0.0006%	0.0007%	0.0011%	0.0009%	0.0007%	0.0009%	0.0008%	0.0010%
9 PM	0.0183%	0.0005%	0.0007%		0.0006%	0.0007%	0.0008%	0.0011%	0.0010%	0.0008%	0.0009%	0.0008%	0.0009%
10 PM	0.0006%	0.0005%	0.0008%		0.0007%	0.0007%	0.0008%	0.0012%	0.0009%	0.0008%	0.0009%	0.0008%	0.0010%
11 PM	0.0006%	0.0005%	0.0008%		0.0007%	0.0007%	0.0008%	0.0012%	0.0009%	0.0170%	0.0009%	0.0008%	0.0010%

## Activity Distribution Map 3

Hydro Power Map Switzerland													
Hydro	January	February	March	April	May	June	July	August	September	October	November	December	
	8.29%		7.02%	6.83%	5.79%	8.23%	12.13%	12.52%	11.30%	7.86%	6.08%	7.28%	6.68%

## Activity Distribution Map 4

Nuclear Activity Distribution MAP Switzerland													
Nuclear	January	February	March	April	May	June	July	August	September	October	November	December	
	11.6%	10.5%	11.7%		11.4%	8.5%	3.4%	6.8%	5.6%	6.0%	6.9%	6.6%	11.0%

## 0.1 Cost Minimization Results

Table 0.1.1 Oil Utilization

Time	Oil GW MAP Switzerland											
	January	February	March	April	May	June	July	August	September	October	November	December
12 AM	0.002053388	0.002053388	0.0020534	0.00205	0.002053	0.00205	0.0020534	0.00205339	0.002053388	0.002053388	0.002053388	0.002053388
1 AM	0.002053388	0.002053388	0.0020534	0.00205	0.002053	0.00205	0.0020534	0.00205339	0.002053388	0.002053388	0.002053388	0.002053388
2 AM	0.002053388	0.002053388	0.0020534	0.00205	0.002053	0.00205	0.0020534	0.00205339	0.002053388	0.002053388	0.002053388	0.002053388
3 AM	0.002053388	0.002053388	0.0020534	0.00205	0.002053	0.00205	0.0020534	0.00205339	0.002053388	0.002053388	0.002053388	0.002053388
4 AM	0.002053388	0.002053388	0.0020534	0.00205	0.002053	0.00205	0.0020534	0.00205339	0.002053388	0.002053388	0.002053388	0.002053388
5 AM	0.002053388	0.002053388	0.0020534	0.00205	0.002053	0.00205	0.0020534	0.00205339	0.002053388	0.002053388	0.002053388	0.002053388
6 AM	0.002053388	0.002053388	0.0020534	0.00205	0.002053	0.00205	0.0020534	0.00205339	0.002053388	0.002053388	0.002053388	0.002053388
7 AM	0.002053388	0.002053388	0.0020534	0.00205	0.002053	0.00205	0.0020534	0.00205339	0.002053388	0.002053388	0.002053388	0.002053388
8 AM	0.002053388	0.002053388	0.0020534	0.00205	0.002053	0.00205	0.0020534	0.00205339	0.002053388	0.002053388	0.002053388	0.002053388
9 AM	0.002053388	0.002053388	0.0020534	0.00205	0.002053	0.00205	0.0020534	0.00205339	0.002053388	0.002053388	0.002053388	0.002053388
10 AM	0.002053388	0.002053388	0.0020534	0.00205	0.002053	0.00205	0.0020534	0.00205339	0.002053388	0.002053388	0.002053388	0.002053388
11 AM	0.002053388	0.002053388	0.0020534	0.00205	0.002053	0.00205	0.0020534	0.00205339	0.002053388	0.002053388	0.002053388	0.002053388
12 PM	0.002053388	0.002053388	0.0020534	0.00205	0.002053	0.00205	0.0020534	0.00205339	0.002053388	0.002053388	0.002053388	0.002053388
1 PM	0.002053388	0.002053388	0.0020534	0.00205	0.002053	0.00205	0.0020534	0.00205339	0.002053388	0.002053388	0.002053388	0.002053388
2 PM	0.002053388	0.002053388	0.0020534	0.00205	0.002053	0.00205	0.0020534	0.00205339	0.002053388	0.002053388	0.002053388	0.002053388
3 PM	0.002053388	0.002053388	0.0020534	0.00205	0.002053	0.00205	0.0020534	0.00205339	0.002053388	0.002053388	0.002053388	0.002053388
4 PM	0.002053388	0.002053388	0.0020534	0.00205	0.002053	0.00205	0.0020534	0.00205339	0.002053388	0.002053388	0.002053388	0.002053388
5 PM	0.002053388	0.002053388	0.0020534	0.00205	0.002053	0.00205	0.0020534	0.00205339	0.002053388	0.002053388	0.002053388	0.002053388
6 PM	0.002053388	0.002053388	0.0020534	0.00205	0.002053	0.00205	0.0020534	0.00205339	0.002053388	0.002053388	0.002053388	0.002053388
7 PM	0.002053388	0.002053388	0.0020534	0.00205	0.002053	0.00205	0.0020534	0.00205339	0.002053388	0.002053388	0.002053388	0.002053388
8 PM	0.002053388	0.002053388	0.0020534	0.00205	0.002053	0.00205	0.0020534	0.00205339	0.002053388	0.002053388	0.002053388	0.002053388
9 PM	0.002053388	0.002053388	0.0020534	0.00205	0.002053	0.00205	0.0020534	0.00205339	0.002053388	0.002053388	0.002053388	0.002053388
10 PM	0.002053388	0.002053388	0.0020534	0.00205	0.002053	0.00205	0.0020534	0.00205339	0.002053388	0.002053388	0.002053388	0.002053388
11 PM	0.002053388	0.002053388	0.0020534	0.00205	0.002053	0.00205	0.0020534	0.00205339	0.002053388	0.002053388	0.002053388	0.002053388

Table 0.1.2 Solar Utilization

Time	SOLAR GW MAP Switzerland											
	January	February	March	April	May	June	July	August	September	October	November	December
12 AM	0.002574805	0.00198231	0.0030317	0.00275	0.002932	0.00307	0.004487	0.00374556	0.003261989	0.025957716	0.003394295	0.00387652
1 AM	0.002600877	0.001949186	0.0031741	0.05673	0.002861	0.00297	0.0044825	0.00353444	0.045778358	0.003465028	0.003472975	0.003915915
2 AM	0.002553787	0.001927434	0.0033619	0.00281	0.006219	0.01598	0.0074496	0.00338326	0.003105849	0.003390287	0.003446834	0.003851401
3 AM	0.002571833	0.001930693	0.0035666	0.02581	0.206917	0.32106	0.2266455	0.05121798	0.003493077	0.003320203	0.003325519	0.00383515
4 AM	0.002512748	0.00189044	0.0396091	0.44759	0.958805	1.13321	1.0427982	0.60344298	0.163088033	0.012361096	0.003267257	0.003846879
5 AM	0.00286358	0.035396476	0.4807647	1.12775	1.356289	2.26605	1.0849331	1.42809548	1.030866425	0.320030747	0.028959634	0.004009046
6 AM	0.066864906	0.391845103	1.2216035	1.15666	1.278071	2.1443	1.1028304	1.42611558	2.092467108	1.193380561	0.387523308	0.096606572
7 AM	0.419538113	1.12169212	1.4015119	1.28088	1.279372	2.18451	1.0588003	1.56308934	2.066785346	2.255038269	1.084160785	0.561336815
8 AM	0.927402398	1.869037179	1.5523634	1.33512	1.344119	2.30733	1.2418796	1.76283593	2.318068134	2.817169697	1.783977258	1.166080307
9 AM	1.352862309	2.13793912	1.6254594	1.25805	1.350782	2.34156	1.343276	1.86740499	2.435960581	3.005850772	2.250814842	1.626677279
10 AM	1.543445759	2.069325905	1.4478059	1.05612	1.308389	2.22611	1.2848038	1.77892551	2.327623732	3.019836309	2.359481218	1.785027005
11 AM	1.448805718	2.012350872	1.3893445	0.99141	1.391491	2.27285	1.3843595	1.79632076	2.315866465	2.966350079	2.126075691	1.587575394
12 PM	1.118458155	2.052497179	1.4700465	1.01272	1.466545	2.36677	1.4448963	1.83623585	2.381869025	2.911340588	1.548956199	1.11987451
1 PM	0.627553069	1.517537563	1.3165983	0.83732	1.333489	2.19645	1.2264296	1.68046119	2.123638582	1.889501113	0.779287257	0.507975296
2 PM	0.162608181	0.708110567	1.3070239	0.93605	1.463404	2.35104	1.3810539	1.8391019	2.116181141	0.798546382	0.154593026	0.06799186
3 PM	0.00645051	0.124263098	0.8349498	0.94366	1.549896	2.23357	1.4072702	1.74892699	0.805842158	0.118792795	0.005790521	0.004424831
4 PM	0.002976595	0.002810638	0.0685796	0.37466	0.74994	1.07908	1.1261652	0.61520872	0.103556561	0.004751261	0.004193407	0.004424745
5 PM	0.002824446	0.002276263	0.0030277	0.01775	0.345602	0.31543	0.3034829	0.07033836	0.003176976	0.00361697	0.003846706	0.004281541
6 PM	0.002716764	0.00215284	0.0027497	0.00207	0.002973	0.01662	0.0133294	0.00333757	0.002837527	0.003405595	0.003635294	0.004108511
7 PM	0.002684897	0.002063175	0.0026268	0.00206	0.001907	0.00238	0.00374	0.00335058	0.002750516	0.003347435	0.003325189	0.003989581
8 PM	0.002663034	0.002022047	0.0027044	0.00224	0.002202	0.00273	0.0041831	0.00359006	0.002953456	0.003423455	0.003187968	0.003765927
9 PM	0.072296565	0.002004703	0.0028619	0.0025	0.002708	0.0031	0.0045317	0.00377418	0.003165868	0.003593663	0.003194999	0.003685885
10 PM	0.002533709	0.001946661	0.0029763	0.00264	0.002898	0.0031	0.0045997	0.00373328	0.003245157	0.003684267	0.003231102	0.003749182
11 PM	0.002544267	0.001910278	0.0029942	0.00274	0.002886	0.00307	0.0045914	0.00373281	0.067132002	0.00371617	0.003344379	0.003774478



Table 0.1.3 Wind Utilization

Time	WIND GW MAP Switzerland											
	January	February	March	April	May	June	July	August	September	October	November	December
12 AM	0.576821029	0.618341422	0.6059015	0.49785	0.468764	0.28652	0.3065133	0.30529292	0.365509635	0.477571338	0.571125557	0.470838712
1 AM	0.570797021	0.606407401	0.5958504	0.49093	0.449978	0.28599	0.3127464	0.30811837	0.355457463	0.468239125	0.566655885	0.466726291
2 AM	0.567201499	0.598272446	0.5834811	0.50597	0.443033	0.27842	0.2953607	0.29603583	0.363903994	0.457157376	0.565471981	0.479782423
3 AM	0.553750401	0.606993333	0.5858461	0.5145	0.41887	0.28421	0.3006877	0.30621274	0.368258003	0.461122957	0.550112172	0.470125974
4 AM	0.5720935	0.607316432	0.5949093	0.51397	0.390622	0.25743	0.2956617	0.30825187	0.364671014	0.455735132	0.550542301	0.473814329
5 AM	0.562715093	0.600872864	0.5916107	0.46644	0.34675	0.22398	0.2642817	0.28848945	0.341856594	0.439033064	0.546918725	0.488780802
6 AM	0.571405665	0.568497104	0.5720458	0.41317	0.323881	0.21497	0.2358247	0.25694907	0.317522821	0.394236132	0.540683949	0.48457386
7 AM	0.558753065	0.53029384	0.5242151	0.37992	0.327768	0.20213	0.2287296	0.24426956	0.284736222	0.354370033	0.516659135	0.45529183
8 AM	0.551195515	0.514479282	0.4949703	0.37384	0.341842	0.23516	0.2678976	0.23397743	0.276114886	0.328460755	0.487543584	0.417771675
9 AM	0.539926705	0.51173545	0.5022126	0.40946	0.388575	0.30069	0.3428705	0.27388046	0.3113485	0.318697821	0.448321571	0.369182369
10 AM	0.531079276	0.505814756	0.5372921	0.44542	0.428836	0.37037	0.438947	0.34889553	0.380867477	0.322645508	0.433744912	0.352682104
11 AM	0.54632407	0.512045715	0.6000871	0.47739	0.470913	0.42363	0.4932367	0.41032245	0.446452214	0.361193934	0.460725553	0.438552012
12 PM	0.556322547	0.526789414	0.6331153	0.48931	0.504448	0.44117	0.5053994	0.44731373	0.486912291	0.384656323	0.471745253	0.333078492
1 PM	0.568495948	0.5217638	0.6500599	0.50669	0.506543	0.43312	0.5009339	0.4467325	0.492729657	0.395334509	0.49335977	0.330588763
2 PM	0.591455913	0.535907618	0.6480172	0.50021	0.520695	0.42249	0.486037	0.44485884	0.488467969	0.399251105	0.50305184	0.439945969
3 PM	0.607905563	0.556678595	0.6309995	0.49644	0.527179	0.41048	0.4706042	0.41001483	0.456297189	0.412837993	0.537080306	0.374818873
4 PM	0.617055031	0.590207245	0.6296782	0.47632	0.483223	0.3693	0.4295899	0.37844348	0.435270559	0.445595538	0.543158766	0.414638694
5 PM	0.624262825	0.519244314	0.6455395	0.44137	0.486773	0.33183	0.3808938	0.35818916	0.411273284	0.471814468	0.55504397	0.439916619
6 PM	0.633929057	0.627748127	0.661849	0.42575	0.500106	0.31779	0.3494971	0.32758484	0.398890415	0.479304503	0.567621577	0.46364849
7 PM	0.633728037	0.622014373	0.658654	0.45332	0.514289	0.32333	0.3192968	0.33230374	0.40016205	0.461785791	0.586533624	0.469126098
8 PM	0.643825244	0.619244314	0.6455395	0.47292	0.510498	0.32423	0.3151729	0.3263508	0.399578884	0.466976276	0.548461317	0.493802789
9 PM	0.644898875	0.623188467	0.6527681	0.48357	0.494419	0.29733	0.3111514	0.31971612	0.39575611	0.487047337	0.605501798	0.503946535
10 PM	0.632803648	0.615869351	0.6462903	0.47774	0.482113	0.28687	0.3276402	0.32422943	0.397490397	0.503983167	0.606786415	0.494540589
11 PM	0.610356894	0.612064705	0.634157	0.48244	0.467433	0.28346	0.3186217	0.31463122	0.38532115	0.469786118	0.598890397	0.486881163

Table 0.1.4 Biofuels Utilization

Time	Biofuels GW MAP Switzerland											
	January	February	March	April	May	June	July	August	September	October	November	December
12 AM	0.079055441	0.079055441	0.0790554	0.07906	0.079055	0.07906	0.0790554	0.07905544	0.079055441	0.079055441	0.079055441	0.079055441
1 AM	0.079055441	0.079055441	0.0790554	0.07906	0.079055	0.07906	0.0790554	0.07905544	0.079055441	0.079055441	0.079055441	0.079055441
2 AM	0.079055441	0.079055441	0.0790554	0.07906	0.079055	0.07906	0.0790554	0.07905544	0.079055441	0.079055441	0.079055441	0.079055441
3 AM	0.079055441	0.079055441	0.0790554	0.07906	0.079055	0.07906	0.0790554	0.07905544	0.079055441	0.079055441	0.079055441	0.079055441
4 AM	0.079055441	0.079055441	0.0790554	0.07906	0.079055	0.07906	0.0790554	0.07905544	0.079055441	0.079055441	0.079055441	0.079055441
5 AM	0.079055441	0.079055441	0.0790554	0.07906	0.079055	0.07906	0.0790554	0.07905544	0.079055441	0.079055441	0.079055441	0.079055441
6 AM	0.079055441	0.079055441	0.0790554	0.07906	0.079055	0.07906	0.0790554	0.07905544	0.079055441	0.079055441	0.079055441	0.079055441
7 AM	0.079055441	0.079055441	0.0790554	0.07906	0.079055	0.07906	0.0790554	0.07905544	0.079055441	0.079055441	0.079055441	0.079055441
8 AM	0.079055441	0.079055441	0.0790554	0.07906	0.079055	0.07906	0.0790554	0.07905544	0.079055441	0.079055441	0.079055441	0.079055441
9 AM	0.079055441	0.079055441	0.0790554	0.07906	0.079055	0.07906	0.0790554	0.07905544	0.079055441	0.079055441	0.079055441	0.079055441
10 AM	0.079055441	0.079055441	0.0790554	0.07906	0.079055	0.07906	0.0790554	0.07905544	0.079055441	0.079055441	0.079055441	0.079055441
11 AM	0.079055441	0.079055441	0.0790554	0.07906	0.079055	0.07906	0.0790554	0.07905544	0.079055441	0.079055441	0.079055441	0.079055441
12 PM	0.079055441	0.079055441	0.0790554	0.07906	0.079055	0.07906	0.0790554	0.07905544	0.079055441	0.079055441	0.079055441	0.079055441
1 PM	0.079055441	0.079055441	0.0790554	0.07906	0.079055	0.07906	0.0790554	0.07905544	0.079055441	0.079055441	0.079055441	0.079055441
2 PM	0.079055441	0.079055441	0.0790554	0.07906	0.079055	0.07906	0.0790554	0.07905544	0.079055441	0.079055441	0.079055441	0.079055441
3 PM	0.079055441	0.079055441	0.0790554	0.07906	0.079055	0.07906	0.0790554	0.07905544	0.079055441	0.079055441	0.079055441	0.079055441
4 PM	0.079055441	0.079055441	0.0790554	0.07906	0.079055	0.07906	0.0790554	0.07905544	0.079055441	0.079055441	0.079055441	0.079055441
5 PM	0.079055441	0.079055441	0.0790554	0.07906	0.079055	0.07906	0.0790554	0.07905544	0.079055441	0.079055441	0.079055441	0.079055441
6 PM	0.079055441	0.079055441	0.0790554	0.07906	0.079055	0.07906	0.0790554	0.07905544	0.079055441	0.079055441	0.079055441	0.079055441
7 PM	0.079055441	0.079055441	0.0790554	0.07906	0.079055	0.07906	0.0790554	0.07905544	0.079055441	0.079055441	0.079055441	0.079055441
8 PM	0.079055441	0.079055441	0.0790554	0.07906	0.079055	0.07906	0.0790554	0.07905544	0.079055441	0.079055441	0.079055441	0.079055441
9 PM	0.079055441	0.079055441	0.0790554	0.07906	0.079055	0.07906	0.0790554	0.07905544	0.079055441	0.079055441	0.079055441	0.079055441
10 PM	0.079055441	0.079055441	0.0790554	0.07906	0.079055	0.07906	0.0790554	0.07905544	0.079055441	0.079055441	0.079055441	0.079055441
11 PM	0.079055441	0.079055441	0.0790554	0.07906	0.079055	0.07906	0.0790554	0.07905544	0.079055441	0.079055441	0.079055441	0.079055441

Table 0.1.5 Natural Gas Utilization

	Natural Gas GW MAP Switzerland											
Time	January	February	March	April	May	June	July	August	September	October	November	December
12 AM	0.061031257	0.061031257	0.0610313	0.06103	0.061031	0.06103	0.0610313	0.06103126	0.061031257	0.061031257	0.061031257	0.061031257
1 AM	0.061031257	0.061031257	0.0610313	0.06103	0.061031	0.06103	0.0610313	0.06103126	0.061031257	0.061031257	0.061031257	0.061031257
2 AM	0.061031257	0.061031257	0.0610313	0.06103	0.061031	0.06103	0.0610313	0.06103126	0.061031257	0.061031257	0.061031257	0.061031257
3 AM	0.061031257	0.061031257	0.0610313	0.06103	0.061031	0.06103	0.0610313	0.06103126	0.061031257	0.061031257	0.061031257	0.061031257
4 AM	0.061031257	0.061031257	0.0610313	0.06103	0.061031	0.06103	0.0610313	0.06103126	0.061031257	0.061031257	0.061031257	0.061031257
5 AM	0.061031257	0.061031257	0.0610313	0.06103	0.061031	0.06103	0.0610313	0.06103126	0.061031257	0.061031257	0.061031257	0.061031257
6 AM	0.061031257	0.061031257	0.0610313	0.06103	0.061031	0.06103	0.0610313	0.06103126	0.061031257	0.061031257	0.061031257	0.061031257
7 AM	0.061031257	0.061031257	0.0610313	0.06103	0.061031	0.06103	0.0610313	0.06103126	0.061031257	0.061031257	0.061031257	0.061031257
8 AM	0.061031257	0.061031257	0.0610313	0.06103	0.061031	0.06103	0.0610313	0.06103126	0.061031257	0.061031257	0.061031257	0.061031257
9 AM	0.061031257	0.061031257	0.0610313	0.06103	0.061031	0.06103	0.0610313	0.06103126	0.061031257	0.061031257	0.061031257	0.061031257
10 AM	0.061031257	0.061031257	0.0610313	0.06103	0.061031	0.06103	0.0610313	0.06103126	0.061031257	0.061031257	0.061031257	0.061031257
11 AM	0.061031257	0.061031257	0.0610313	0.06103	0.061031	0.06103	0.0610313	0.06103126	0.061031257	0.061031257	0.061031257	0.061031257
12 PM	0.061031257	0.061031257	0.0610313	0.06103	0.061031	0.06103	0.0610313	0.06103126	0.061031257	0.061031257	0.061031257	0.061031257
1 PM	0.061031257	0.061031257	0.0610313	0.06103	0.061031	0.06103	0.0610313	0.06103126	0.061031257	0.061031257	0.061031257	0.061031257
2 PM	0.061031257	0.061031257	0.0610313	0.06103	0.061031	0.06103	0.0610313	0.06103126	0.061031257	0.061031257	0.061031257	0.061031257
3 PM	0.061031257	0.061031257	0.0610313	0.06103	0.061031	0.06103	0.0610313	0.06103126	0.061031257	0.061031257	0.061031257	0.061031257
4 PM	0.061031257	0.061031257	0.0610313	0.06103	0.061031	0.06103	0.0610313	0.06103126	0.061031257	0.061031257	0.061031257	0.061031257
5 PM	0.061031257	0.061031257	0.0610313	0.06103	0.061031	0.06103	0.0610313	0.06103126	0.061031257	0.061031257	0.061031257	0.061031257
6 PM	0.061031257	0.061031257	0.0610313	0.06103	0.061031	0.06103	0.0610313	0.06103126	0.061031257	0.061031257	0.061031257	0.061031257
7 PM	0.061031257	0.061031257	0.0610313	0.06103	0.061031	0.06103	0.0610313	0.06103126	0.061031257	0.061031257	0.061031257	0.061031257
8 PM	0.061031257	0.061031257	0.0610313	0.06103	0.061031	0.06103	0.0610313	0.06103126	0.061031257	0.061031257	0.061031257	0.061031257
9 PM	0.061031257	0.061031257	0.0610313	0.06103	0.061031	0.06103	0.0610313	0.06103126	0.061031257	0.061031257	0.061031257	0.061031257
10 PM	0.061031257	0.061031257	0.0610313	0.06103	0.061031	0.06103	0.0610313	0.06103126	0.061031257	0.061031257	0.061031257	0.061031257
11 PM	0.061031257	0.061031257	0.0610313	0.06103	0.061031	0.06103	0.0610313	0.06103126	0.061031257	0.061031257	0.061031257	0.061031257

Table 0.1.6 Waste Utilization

Time	Waste GW MAP Switzerland											
	January	February	March	April	May	June	July	August	September	October	November	December
12 AM	0.237394479	0.237394479	0.2373945	0.23739	0.237394	0.23739	0.2373945	0.23739448	0.237394479	0.237394479	0.237394479	0.237394479
1 AM	0.237394479	0.237394479	0.2373945	0.23739	0.237394	0.23739	0.2373945	0.23739448	0.237394479	0.237394479	0.237394479	0.237394479
2 AM	0.237394479	0.237394479	0.2373945	0.23739	0.237394	0.23739	0.2373945	0.23739448	0.237394479	0.237394479	0.237394479	0.237394479
3 AM	0.237394479	0.237394479	0.2373945	0.23739	0.237394	0.23739	0.2373945	0.23739448	0.237394479	0.237394479	0.237394479	0.237394479
4 AM	0.237394479	0.237394479	0.2373945	0.23739	0.237394	0.23739	0.2373945	0.23739448	0.237394479	0.237394479	0.237394479	0.237394479
5 AM	0.237394479	0.237394479	0.2373945	0.23739	0.237394	0.23739	0.2373945	0.23739448	0.237394479	0.237394479	0.237394479	0.237394479
6 AM	0.237394479	0.237394479	0.2373945	0.23739	0.237394	0.23739	0.2373945	0.23739448	0.237394479	0.237394479	0.237394479	0.237394479
7 AM	0.237394479	0.237394479	0.2373945	0.23739	0.237394	0.23739	0.2373945	0.23739448	0.237394479	0.237394479	0.237394479	0.237394479
8 AM	0.237394479	0.237394479	0.2373945	0.23739	0.237394	0.23739	0.2373945	0.23739448	0.237394479	0.237394479	0.237394479	0.237394479
9 AM	0.237394479	0.237394479	0.2373945	0.23739	0.237394	0.23739	0.2373945	0.23739448	0.237394479	0.237394479	0.237394479	0.237394479
10 AM	0.237394479	0.237394479	0.2373945	0.23739	0.237394	0.23739	0.2373945	0.23739448	0.237394479	0.237394479	0.237394479	0.237394479
11 AM	0.237394479	0.237394479	0.2373945	0.23739	0.237394	0.23739	0.2373945	0.23739448	0.237394479	0.237394479	0.237394479	0.237394479
12 PM	0.237394479	0.237394479	0.2373945	0.23739	0.237394	0.23739	0.2373945	0.23739448	0.237394479	0.237394479	0.237394479	0.237394479
1 PM	0.237394479	0.237394479	0.2373945	0.23739	0.237394	0.23739	0.2373945	0.23739448	0.237394479	0.237394479	0.237394479	0.237394479
2 PM	0.237394479	0.237394479	0.2373945	0.23739	0.237394	0.23739	0.2373945	0.23739448	0.237394479	0.237394479	0.237394479	0.237394479
3 PM	0.237394479	0.237394479	0.2373945	0.23739	0.237394	0.23739	0.2373945	0.23739448	0.237394479	0.237394479	0.237394479	0.237394479
4 PM	0.237394479	0.237394479	0.2373945	0.23739	0.237394	0.23739	0.2373945	0.23739448	0.237394479	0.237394479	0.237394479	0.237394479
5 PM	0.237394479	0.237394479	0.2373945	0.23739	0.237394	0.23739	0.2373945	0.23739448	0.237394479	0.237394479	0.237394479	0.237394479
6 PM	0.237394479	0.237394479	0.2373945	0.23739	0.237394	0.23739	0.2373945	0.23739448	0.237394479	0.237394479	0.237394479	0.237394479
7 PM	0.237394479	0.237394479	0.2373945	0.23739	0.237394	0.23739	0.2373945	0.23739448	0.237394479	0.237394479	0.237394479	0.237394479
8 PM	0.237394479	0.237394479	0.2373945	0.23739	0.237394	0.23739	0.2373945	0.23739448	0.237394479	0.237394479	0.237394479	0.237394479
9 PM	0.237394479	0.237394479	0.2373945	0.23739	0.237394	0.23739	0.2373945	0.23739448	0.237394479	0.237394479	0.237394479	0.237394479
10 PM	0.237394479	0.237394479	0.2373945	0.23739	0.237394	0.23739	0.2373945	0.23739448	0.237394479	0.237394479	0.237394479	0.237394479
11 PM	0.237394479	0.237394479	0.2373945	0.23739	0.237394	0.23739	0.2373945	0.23739448	0.237394479	0.237394479	0.237394479	0.237394479

Table 0.1.7 Nuclear Utilization

Time	Nuclear GW MAP Switzerland											
	January	February	March	April	May	June	July	August	September	October	November	December
12 AM	3.071365339	2.783827177	3.1099899	3.00985	2.254528	0.89981	1.799617	1.48775964	1.600772154	1.826797176	1.75527027	2.908283993
1 AM	3.071365339	2.783827177	3.1099899	3.00985	2.254528	0.89981	1.799617	1.48775964	1.600772154	1.826797176	1.75527027	2.908283993
2 AM	3.071365339	2.783827177	3.1099899	3.00985	2.254528	0.89981	1.799617	1.48775964	1.600772154	1.826797176	1.75527027	2.908283993
3 AM	3.071365339	2.783827177	3.1099899	3.00985	2.254528	0.89981	1.799617	1.48775964	1.600772154	1.826797176	1.75527027	2.908283993
4 AM	3.071365339	2.783827177	3.1099899	3.00985	2.254528	0.89981	1.799617	1.48775964	1.600772154	1.826797176	1.75527027	2.908283993
5 AM	3.071365339	2.783827177	3.1099899	3.00985	2.254528	0.89981	1.799617	1.48775964	1.600772154	1.826797176	1.75527027	2.908283993
6 AM	3.071365339	2.783827177	3.1099899	3.00985	2.254528	0.89981	1.799617	1.48775964	1.600772154	1.826797176	1.75527027	2.908283993
7 AM	3.071365339	2.783827177	3.1099899	3.00985	2.254528	0.89981	1.799617	1.48775964	1.600772154	1.826797176	1.75527027	2.908283993
8 AM	3.071365339	2.783827177	3.1099899	3.00985	2.254528	0.89981	1.799617	1.48775964	1.600772154	1.826797176	1.75527027	2.908283993
9 AM	3.071365339	2.783827177	3.1099899	3.00985	2.254528	0.89981	1.799617	1.48775964	1.600772154	1.826797176	1.75527027	2.908283993
10 AM	3.071365339	2.783827177	3.1099899	3.00985	2.254528	0.89981	1.799617	1.48775964	1.600772154	1.826797176	1.75527027	2.908283993
11 AM	3.071365339	2.783827177	3.1099899	3.00985	2.254528	0.89981	1.799617	1.48775964	1.600772154	1.826797176	1.75527027	2.908283993
12 PM	3.071365339	2.783827177	3.1099899	3.00985	2.254528	0.89981	1.799617	1.48775964	1.600772154	1.826797176	1.75527027	2.908283993
1 PM	3.071365339	2.783827177	3.1099899	3.00985	2.254528	0.89981	1.799617	1.48775964	1.600772154	1.826797176	1.75527027	2.908283993
2 PM	3.071365339	2.783827177	3.1099899	3.00985	2.254528	0.89981	1.799617	1.48775964	1.600772154	1.826797176	1.75527027	2.908283993
3 PM	3.071365339	2.783827177	3.1099899	3.00985	2.254528	0.89981	1.799617	1.48775964	1.600772154	1.826797176	1.75527027	2.908283993
4 PM	3.071365339	2.783827177	3.1099899	3.00985	2.254528	0.89981	1.799617	1.48775964	1.600772154	1.826797176	1.75527027	2.908283993
5 PM	3.071365339	2.783827177	3.1099899	3.00985	2.254528	0.89981	1.799617	1.48775964	1.600772154	1.826797176	1.75527027	2.908283993
6 PM	3.071365339	2.783827177	3.1099899	3.00985	2.254528	0.89981	1.799617	1.48775964	1.600772154	1.826797176	1.75527027	2.908283993
7 PM	3.071365339	2.783827177	3.1099899	3.00985	2.254528	0.89981	1.799617	1.48775964	1.600772154	1.826797176	1.75527027	2.908283993
8 PM	3.071365339	2.783827177	3.1099899	3.00985	2.254528	0.89981	1.799617	1.48775964	1.600772154	1.826797176	1.75527027	2.908283993
9 PM	3.071365339	2.783827177	3.1099899	3.00985	2.254528	0.89981	1.799617	1.48775964	1.600772154	1.826797176	1.75527027	2.908283993
10 PM	3.071365339	2.783827177	3.1099899	3.00985	2.254528	0.89981	1.799617	1.48775964	1.600772154	1.826797176	1.75527027	2.908283993
11 PM	3.071365339	2.783827177	3.1099899	3.00985	2.254528	0.89981	1.799617	1.48775964	1.600772154	1.826797176	1.75527027	2.908283993

Table 0.1.8 Hydro Powerplant Utilization

	Hydro Power Map Switzerland											
Time	January	February	March	April	May	June	July	August	September	October	November	December
12 AM	2.07911466	1.651280923	1.0484641	0.711124	0.688876	1.74116	0.7263198	1.30838901	1.843993777	2.253462122	3.393552037	2.782493007
1 AM	2.108078107	1.631501614	0.9639689	0.69912	0.835856	1.8269	0.8290299	1.51988102	1.828863186	2.282439592	3.249055204	2.631243601
2 AM	2.198848023	1.760838242	1.0931478	0.92576	1.1391	2.0222	1.0227962	1.63255616	2.016600418	2.396132507	3.412022137	2.593790895
3 AM	2.314607437	1.939853541	1.2110118	1.02969	1.133967	1.92764	0.9327516	1.63946737	2.161573844	2.50043041	3.540515028	2.702495416
4 AM	2.376291882	1.961268847	1.1791288	0.71236	0.449203	1.22634	0.1769911	0.97980362	2.005489602	2.514095247	3.629085516	2.795983132
5 AM	2.280847558	1.831741056	0.7015573	0	0	0	0	0	1.019535138	2.162066027	3.47571605	2.714684302
6 AM	2.134173029	1.434204661	0	0	0	0	0	0	0	1.42864565	3.122577033	2.714660623
7 AM	1.709474551	0.733816506	0	0	0	0	0	0	0	0.372145989	2.443852888	2.095207783
8 AM	1.419324709	0.172829143	0	0	0	0	0	0	0	0	2.00593127	1.509208289
9 AM	1.138644607	0	0	0	0	0	0	0	0	0	1.739829423	1.221490108
10 AM	0.950956522	0	0	0	0	0	0	0	0	0	1.642725666	1.106837788
11 AM	1.044337407	0	0	0	0	0	0	0	0	0	1.885481055	1.385391618
12 PM	1.406043771	0	0	0	0	0	0	0	0	0.017288656	2.532226763	1.947302229
1 PM	1.867582679	0.44842889	0	0	0	0	0	0	0	0.76470548	3.108296204	2.554226456
2 PM	2.359840064	1.278175509	0	0	0	0	0	0	0.051995254	1.895341447	3.814581028	2.996601441
3 PM	2.423716782	1.756082609	0.4193029	0	0	0.1463	0	0.15939035	1.455567793	2.491417202	3.870576938	3.004097512
4 PM	2.350061098	1.798908916	1.1501529	0.60797	0.913232	1.28925	0.3861403	1.30687346	2.295979223	2.526779139	3.849327316	2.949774486
5 PM	2.281064194	1.73848871	1.2527578	1.03711	1.296881	2.01741	1.1984364	1.83394479	2.398178054	2.516253142	3.852731898	2.895743304
6 PM	2.309158769	1.713804672	1.1927195	1.14399	1.443112	2.24138	1.269946	1.76941333	2.217543881	2.489309301	3.864600652	2.872004515
7 PM	2.341027501	1.955050889	1.379897	1.02603	1.137087	2.25003	1.2299383	1.62519199	2.155913523	2.707102555	3.769240872	2.749407754
8 PM	2.032553442	1.742740598	1.3226058	0.83264	0.851308	2.17091	1.0816841	1.62528797	2.118426968	2.669892767	3.467349914	2.462200751
9 PM	1.797540455	1.574383928	1.080928	0.81256	0.798959	2.06835	0.956168	1.46213159	2.022262753	2.355041649	3.257920037	2.376253087
10 PM	1.791752513	1.465828944	0.9311838	0.75046	0.814312	2.00729	0.9139269	1.40122028	1.808583039	2.254641953	3.210285184	2.427110916
11 PM	1.965567517	1.502242261	0.9868128	0.73433	0.748694	1.96164	0.8720913	1.348849716	1.752378877	2.348849716	3.415339598	2.578067843



Table 0.2.4 Solar Utilization

SOLAR GW MAP Switzerland												
Time	January	February	March	April	May	June	July	August	September	October	November	December
12 AM		0	0	0	0	0	0	0	0	0	0	0
1 AM		0	0	0	0	0	0	0	0	0	0	0
2 AM		0	0	0	0	0	0	0	0	0	0	0
3 AM		0	0	0	0	0	0	0	0	0	0	0
4 AM		0	0	0	0	0	0	0	0	0	0	0
5 AM		0	0	0	0	0	0	0	0	0	0	0
6 AM		0	0	0	0	0	0	0	0	0	0	0
7 AM		0	0	0	0	0	0	0	0	0	0	0
8 AM		0	0	0	0	0	0	0	0	0	0	0
9 AM		0	0	0	0	0	0	0	0	0	0	0
10 AM		0	0	0	0	0	0	0	0	0	0	0
11 AM		0	0	0	0	0	0	0	0	0	0	0
12 PM		0	0	0	0	0	0	0	0	0	0	0
1 PM		0	0	0	0	0	0	0	0	0	0	0
2 PM		0	0	0	0	0	0	0	0	0	0	0
3 PM		0	0	0	0	0	0	0	0	0	0	0
4 PM		0	0	0	0	0	0	0	0	0	0	0
5 PM		0	0	0	0	0	0	0	0	0	0	0
6 PM		0	0	0	0	0	0	0	0	0	0	0
7 PM		0	0	0	0	0	0	0	0	0	0	0
8 PM		0	0	0	0	0	0	0	0	0	0	0
9 PM		0	0	0	0	0	0	0	0	0	0	0
10 PM		0	0	0	0	0	0	0	0	0	0	0
11 PM		0	0	0	0	0	0	0	0	0	0	0

Table 0.2.5 Hydro Utilization

Hydro Power Map Switzerland												
Time	January	February	March	April	May	June	July	August	September	October	November	December
12 AM	2.46122403	2.032797798	1.4310304	1.09352451	1.0713429	2.12375949	1.11034141	1.6916691	2.226790332	2.658954404	3.776480898	3.165904093
1 AM	2.490213549	2.012985365	1.3466776	1.13538333	1.2182515	2.20940262	1.21304699	1.90295	2.254176109	2.665439185	3.632062744	3.014694081
2 AM	2.580936376	2.142300241	1.4760443	1.30810498	1.5248528	2.41771392	1.40978038	2.015474	2.399240833	2.779057359	3.795003536	2.977176862
3 AM	2.696713835	2.321318799	1.5941129	1.43503475	1.7204184	2.62822683	1.53893171	2.0702199	2.544601486	2.883285178	3.923375112	3.085865131
4 AM	2.758339195	2.342693853	1.5982725	1.53948684	1.7875422	2.73909156	1.59932379	1.9627812	2.548112201	2.905990908	4.011887338	3.179364576
5 AM	2.663245703	2.246672097	1.5618566	1.50728875	1.7358238	2.64558488	1.46446769	1.80763	2.429936129	2.861631339	3.884265804	3.098227914
6 AM	2.580572501	2.20558433	1.601138	1.53619143	1.6576058	2.52383342	1.482365	1.8056501	2.472001674	3.001560776	3.889634907	3.19080176
7 AM	2.50854723	2.235043191	1.7810465	1.66041633	1.6589062	2.56404278	1.43833489	1.9426239	2.446319912	3.006718824	3.907548238	3.036079164
8 AM	2.726261672	2.421400887	1.931898	1.71465101	1.7236538	2.68686088	1.62141416	2.1423705	2.697602699	3.196704262	4.169443094	3.054823161
9 AM	2.871041482	2.517473685	2.004994	1.63758936	1.7303169	2.72109772	1.72281054	2.2469396	2.815495146	3.385385337	4.37017883	3.227701952
10 AM	2.873936847	2.44886047	1.8273404	1.43564997	1.6879239	2.60564784	1.66433835	2.1584601	2.707158298	3.399370874	4.381741449	3.271399359
11 AM	2.872677691	2.391885438	1.7688791	1.37093997	1.771026	2.65238396	1.76389406	2.1758553	2.695401031	3.345884644	4.391091311	3.352501578
12 PM	2.904036492	2.432031745	1.8495811	1.39225603	1.8460795	2.74630059	1.8244309	2.2157704	2.76140359	3.308163809	4.460717527	3.446711305
1 PM	2.874670314	2.345501018	1.6961329	1.21685777	1.7130236	2.57598451	1.60596413	2.0599958	2.503173148	3.033741158	4.267118027	3.441736317
2 PM	2.90198281	2.365820641	1.6865585	1.31558507	1.8429388	2.73056974	1.76058842	2.2186365	2.54771096	3.073422394	4.34870862	3.444127867
3 PM	2.809701858	2.259880272	1.6337873	1.32319253	1.9294306	2.75941281	1.78680477	2.2878519	2.640944516	2.989744563	4.255902024	3.388056908
4 PM	2.732572258	2.181254119	1.598267	1.36216586	2.0427065	2.74786317	1.8918401	2.3016168	2.77907035	2.911064965	4.233055288	3.33373796
5 PM	2.663423205	2.120299539	1.6353201	1.4349837	2.0220178	2.71237022	1.8814539	2.2838177	2.780889595	2.899404678	4.23611317	3.279559411
6 PM	2.691410098	2.095492078	1.5750038	1.52559215	1.8256197	2.63753758	1.66280995	2.1522855	2.599915973	2.872249461	4.247770511	3.255647591
7 PM	2.723246963	2.336648629	1.7620583	1.4076221	1.5185282	2.63193989	1.6132129	2.0080771	2.538198604	3.089984555	4.152100626	3.1329319
8 PM	2.414751041	2.12429721	1.7048447	1.2144143	1.233045	2.55317244	1.46540175	1.9084126	2.50091499	3.052850788	3.850072447	2.845501244
9 PM	2.249371585	1.955923197	1.4633245	1.19458981	1.1812018	2.45098864	1.3402342	1.8454403	2.404963186	2.738169878	3.640649601	2.759473537
10 PM	2.173820787	1.84731017	1.3136947	1.13263524	1.1967443	2.38993217	1.29806117	1.7844881	2.191362762	2.637860786	3.593050852	2.810394663
11 PM	2.347735549	1.883687104	1.3693415	1.11661023	1.1311151	2.34424527	1.25621733	1.7525158	2.199045444	2.732100451	3.798218542	2.961376887

Table 0.2.6 Nuclear Utilization

Nuclear GW MAP Switzerland												
Time	January	February	March	April	May	June	July	August	September	October	November	December
12 AM	3.071365339	2.783827177	3.1099899	3.0098522	2.2545281	0.89980848	1.79961695	1.4877596	1.600772154	1.826797176	1.75527027	2.908283993
1 AM	3.071365339	2.783827177	3.1099899	3.0098522	2.2545281	0.89980848	1.79961695	1.4877596	1.600772154	1.826797176	1.75527027	2.908283993
2 AM	3.071365339	2.783827177	3.1099899	3.0098522	2.2545281	0.89980848	1.79961695	1.4877596	1.600772154	1.826797176	1.75527027	2.908283993
3 AM	3.071365339	2.783827177	3.1099899	3.0098522	2.2545281	0.89980848	1.79961695	1.4877596	1.600772154	1.826797176	1.75527027	2.908283993
4 AM	3.071365339	2.783827177	3.1099899	3.0098522	2.2545281	0.89980848	1.79961695	1.4877596	1.600772154	1.826797176	1.75527027	2.908283993
5 AM	3.071365339	2.783827177	3.1099899	3.0098522	2.2545281	0.89980848	1.79961695	1.4877596	1.600772154	1.826797176	1.75527027	2.908283993
6 AM	3.071365339	2.783827177	3.1099899	3.0098522	2.2545281	0.89980848	1.79961695	1.4877596	1.600772154	1.826797176	1.75527027	2.908283993
7 AM	3.071365339	2.783827177	3.1099899	3.0098522	2.2545281	0.89980848	1.79961695	1.4877596	1.600772154	1.826797176	1.75527027	2.908283993
8 AM	3.071365339	2.783827177	3.1099899	3.0098522	2.2545281	0.89980848	1.79961695	1.4877596	1.600772154	1.826797176	1.75527027	2.908283993
9 AM	3.071365339	2.783827177	3.1099899	3.0098522	2.2545281	0.89980848	1.79961695	1.4877596	1.600772154	1.826797176	1.75527027	2.908283993
10 AM	3.071365339	2.783827177	3.1099899	3.0098522	2.2545281	0.89980848	1.79961695	1.4877596	1.600772154	1.826797176	1.75527027	2.908283993
11 AM	3.071365339	2.783827177	3.1099899	3.0098522	2.2545281	0.89980848	1.79961695	1.4877596	1.600772154	1.826797176	1.75527027	2.908283993
12 PM	3.071365339	2.783827177	3.1099899	3.0098522	2.2545281	0.89980848	1.79961695	1.4877596	1.600772154	1.826797176	1.75527027	2.908283993
1 PM	3.071365339	2.783827177	3.1099899	3.0098522	2.2545281	0.89980848	1.79961695	1.4877596	1.600772154	1.826797176	1.75527027	2.908283993
2 PM	3.071365339	2.783827177	3.1099899	3.0098522	2.2545281	0.89980848	1.79961695	1.4877596	1.600772154	1.826797176	1.75527027	2.908283993
3 PM	3.071365339	2.783827177	3.1099899	3.0098522	2.2545281	0.89980848	1.79961695	1.4877596	1.600772154	1.826797176	1.75527027	2.908283993
4 PM	3.071365339	2.783827177	3.1099899	3.0098522	2.2545281	0.89980848	1.79961695	1.4877596	1.600772154	1.826797176	1.75527027	2.908283993
5 PM	3.071365339	2.783827177	3.1099899	3.0098522	2.2545281	0.89980848	1.79961695	1.4877596	1.600772154	1.826797176	1.75527027	2.908283993
6 PM	3.071365339	2.783827177	3.1099899	3.0098522	2.2545281	0.89980848	1.79961695	1.4877596	1.600772154	1.826797176	1.75527027	2.908283993
7 PM	3.071365339	2.783827177	3.1099899	3.0098522	2.2545281	0.89980848	1.79961695	1.4877596	1.600772154	1.826797176	1.75527027	2.908283993
8 PM	3.071365339	2.783827177	3.1099899	3.0098522	2.2545281	0.89980848	1.79961695	1.4877596	1.600772154	1.826797176	1.75527027	2.908283993
9 PM	3.071365339	2.783827177	3.1099899	3.0098522	2.2545281	0.89980848	1.79961695	1.4877596	1.600772154	1.826797176	1.75527027	2.908283993
10 PM	3.071365339	2.783827177	3.1099899	3.0098522	2.2545281	0.89980848	1.79961695	1.4877596	1.600772154	1.826797176	1.75527027	2.908283993
11 PM	3.071365339	2.783827177	3.1099899	3.0098522	2.2545281	0.89980848	1.79961695	1.4877596	1.600772154	1.826797176	1.75527027	2.908283993

*Table 0.2.7 Waste Utilization*

[illegible]

*Table 0.2.8 Natural Gas Utilization*

[illegible]

Table 0.2.9 Utilization

[illegible]



*Table 0.3 Demand Switzerland*

Row Labels	DEMAND in GWh											
	January	February	March	April	May	June	July	August	September	October	November	December
12 AM	6.109	5.435	5.147	4.601	3.795	3.310	3.216	3.485	4.193	4.963	6.103	6.545
1 AM	6.132	5.403	5.053	4.636	3.923	3.395	3.325	3.699	4.210	4.960	5.954	6.390
2 AM	6.220	5.524	5.170	4.824	4.222	3.596	3.505	3.799	4.364	5.063	6.116	6.365
3 AM	6.322	5.712	5.290	4.959	4.394	3.812	3.639	3.864	4.514	5.171	6.229	6.464
4 AM	6.402	5.734	5.303	5.063	4.433	3.896	3.695	3.759	4.514	5.189	6.318	6.561
5 AM	6.297	5.631	5.263	4.984	4.337	3.769	3.528	3.584	4.373	5.127	6.186	6.495
6 AM	6.223	5.558	5.283	4.959	4.236	3.639	3.518	3.550	4.390	5.223	6.186	6.584
7 AM	6.139	5.549	5.415	5.050	4.241	3.666	3.467	3.675	4.332	5.188	6.179	6.400
8 AM	6.349	5.720	5.537	5.098	4.320	3.822	3.689	3.864	4.574	5.352	6.412	6.381
9 AM	6.482	5.813	5.617	5.057	4.373	3.922	3.865	4.009	4.728	5.531	6.574	6.505
10 AM	6.476	5.739	5.475	4.891	4.371	3.876	3.903	3.995	4.689	5.549	6.571	6.532
11 AM	6.489	5.688	5.479	4.858	4.496	3.976	4.057	4.074	4.743	5.534	6.607	6.599
12 PM	6.532	5.743	5.593	4.891	4.605	4.087	4.129	4.151	4.849	5.520	6.688	6.688
1 PM	6.515	5.651	5.456	4.733	4.474	3.909	3.907	3.994	4.597	5.256	6.516	6.681
2 PM	6.565	5.686	5.445	4.826	4.618	4.055	4.046	4.151	4.637	5.299	6.607	6.702
3 PM	6.489	5.600	5.375	4.829	4.711	4.070	4.057	4.186	4.698	5.229	6.548	6.671
4 PM	6.421	5.555	5.338	4.848	4.780	4.017	4.121	4.168	4.815	5.183	6.531	6.657
5 PM	6.359	5.520	5.391	4.886	4.763	3.944	4.062	4.130	4.793	5.198	6.546	6.628
6 PM	6.397	5.507	5.347	4.961	4.580	3.855	3.812	3.968	4.600	5.178	6.571	6.628
7 PM	6.428	5.742	5.531	4.871	4.287	3.855	3.732	3.828	4.539	5.379	6.494	6.510
8 PM	6.130	5.527	5.460	4.697	3.998	3.777	3.580	3.723	4.501	5.347	6.204	6.247
9 PM	5.966	5.363	5.226	4.688	3.930	3.648	3.451	3.653	4.401	5.052	6.001	6.172
10 PM	5.878	5.247	5.070	4.620	3.933	3.577	3.425	3.596	4.190	4.969	5.955	6.213
11 PM	6.029	5.280	5.113	4.609	3.853	3.528	3.374	3.555	4.185	5.029	6.152	6.357
Grand Total	151.349	133.927	128.377	116.442	103.676	91.000	89.104	92.459	108.428	125.489	152.248	155.975

## 0.3 Supply Chain Risk scores

Aluminium	Fused Al Oxide		Silicon Carbide		Weightd Mining	Governance Index	Weighted index	Recycling Score %:
	2018	2019	2018	2019				
United States	60,000	60,000	40,000	40,000	4.71%	78.99384054	3.717357202	57.3
Argentina	—	0	5,000	5,000	0.00%	42.21330039	0	
Australia	50,000	50,000			3.92%	91.18427531	3.575853934	
Austria	60,000	60,000			4.71%	90.04623795	4.237470021	
Brazil	50,000	50,000	40,000	40,000	3.92%	40.81806405	1.600708394	
China	800,000	800,000	450,000	450,000	62.75%	44.06116374	27.64622039	
France	40,000	40,000	20,000	20,000	3.14%	81.86018372	2.568162626	
Germany	80,000	80,000	35,000	35,000	6.27%	89.43319956	5.611494875	
India	40,000	40,000	5,000	5,000	3.14%	47.7993412	1.499587175	
Japan	15,000	15,000	60,000	60,000	1.18%	88.73260625	1.043913015	
Mexico			45,000	45,000	0.00%	31.93078709	0	
Norway			80,000	80,000	0.00%	96.82661947	0	
Venezuela			30,000	30,000	0.00%	3.261176288	0	
Other	80,000	80,000	190,000	190,000	6.27%			
World	1,300,000	1,300,000	1,000,000	1,000,000			51.50076763	

Chromium	2018	2019	Reserves	Weightd Mining	Governance Index	Weighted index	Recycling Score %:
United States	—	0	620	0%	78.99384054	0.079815164	90.0
Finland	2,210	2,200	13,000	2%	97.00160853	2.402829845	
India	4,300	4,100	100,000	17%	47.7993412	8.109108926	
Kazakhstan	6,690	6,700	230,000	39%	42.04158465	16.21727305	
South Africa	17,600	17,000	200,000	35%	51.35069784	18.15961251	
Turkey	8,000	10,000	26,000	6%	35.23606904	2.067237844	
Other Contries	4,250	4,000		1%			
World total	43,100	44,000	570,000			47.03587734	

Cobalt	2018	2019	Reserves	Weightd Mining	Governance Index	Weighted index	Recycling Score %:
United States	490	500	55,000	0.07%	78.99384054	0.056839679	68.0
Australia	4,880	5,100	71,200,000	92.32%	91.18427531	84.17758442	
Canada	3,520	3,000	230,000	0.30%	91.72706858	0.277088718	
China	2,000	2,000	80,000	0.11%	44.06116374	0.046841978	
Congo Kinsasha	104,000	100,000	3,600,000	4.80%	12.2685914	0.588520824	
Cuba	3,500	3,500	500,000	0.65%	35.31973966	0.230559157	
Madagascar	3,300	3,300	120,000	0.16%	22.70872513	0.036301221	
Morocco	2,100	2,100	18,000	0.03%	40.62635835	0.010586913	
New Caledonia	2,100	1,600	0	0.00%	81.86018372	0.00169808	
Papua New Guinea	3,280	3,100	56,000	0.08%	27.11560822	0.020776493	
Philippines	4,600	4,600	260,000	0.34%	38.35872523	0.131588947	
Russia	6,100	6,100	250,000	0.33%	26.78473822	0.088932887	
South Africa	2,300	2,400	50,000	0.07%	51.35069784	0.034885347	
Other	5,540	5,700	570,000	0.75%			
World	148,000	140,000	76,989,000.00			85.70220467	

Copper	2018	2019	Reserves	Weightd Mining	Governance Index	Weighted index	Recycling Score %:
United States	1,220	1,300	51,000	3%	78.99384054	2.442708823	48.0
Australia	920	960	887,000	53%	91.18427531	47.87294411	
Chile	5,830	5,600	200,000	12%	73.4654541	8.93064983	
China	1,590	1,600	26,000	2%	44.06116374	0.719021421	
Congo	1,230	1,300	19,000	1%	12.2685914	0.147254144	
Indonesia	651	340	28,000	2%	48.91132736	0.819570048	
Kazakhstan	603	700	20,000	1%	42.04158465	0.514548369	
Mexico	751	770	53,000	3%	31.93078709	1.015141176	
Peru	2,440	2,400	87,000	5%	40.80554517	2.156917264	
Russia	751	750	61,000	4%	26.78473822	0.977915099	
Zambia	854	790	19,000	1%	31.76507537	0.371682803	
Other	3,540	3,800	220,000	13%			
World	20,400	20,000	1,671,000			65.96835309	

Graphite	2018	2019	Reserves	Weightd Mining	Governance Index	Weighted index	Recycling Score %:
United States				0.00%	78.99384054	0	0
Austria	1,000	1,000		0.00%	90.04623795	0.00029814	
Brazil	95,000	96,000	72,000,000	23.87%	40.81806405	9.743574436	
Canada	40,000	40,000		0.01%	91.72706858	0.012148209	
China	693,000	700,000	73,000,000	24.40%	44.06116374	10.75172594	
Germany	800	800		0.00%	89.43319956	0.000236888	
India	35,000	35,000	8,000,000	2.66%	47.7993412	1.271635157	
Korea, North	6,000	6,000	2,000,000	0.66%	7.275036256	0.048319321	
Madagascar	46,900	47,000	1,600,000	0.55%	22.70872513	0.123834338	
Mexico	9,000	9,000	3,100,000	1.03%	31.93078709	0.328688932	
Mozambique	104,000	100,000	25,000,000	8.31%	21.38310719	1.77704845	
Namibia	3,460	3,500		0.00%	60.10453097	0.000696514	
Norway	16,000	16,000	600,000	0.20%	96.82661947	0.197483227	
Pakistan	14,000	14,000		0.00%	24.43806529	0.00113279	
Russia	25,200	25,000		0.01%	26.78473822	0.002217084	
Sri Lanka	4,000	4,000		0.00%	44.63527362	0.000591144	
Tanzania	150	150	18,000,000	5.96%	32.13688024	1.91529014	
Turkey	2,000	2,000	90,000,000	29.80%	35.23606904	10.50012205	
Ukraine	20,000	20,000		0.01%	31.89271243	0.002111914	
Vietnam	5,000	5,000	7,600,000	2.52%	42.25464598	1.063967642	
Zimbabwe	2,000	2,000		0.00%	11.34755913	7.51428E-05	
Other	200	200		0.00%			
World	1,120,000	1,100,000	300,000,000			37.74119746	

	Refinery Production					Recycling Score %:
Indium	2018	2019	Weightd Mining	Governance Index	Weighted index	1
Belgium	22	20	2.63%	84.07547251	2.212512435	
Canada	58	60	7.89%	91.72706858	7.241610678	
China	300	300	39.47%	44.06116374	17.39256464	
France	40	50	6.58%	81.86018372	5.385538402	
Japan	70	75	9.87%	88.73260625	8.756507196	
Korea,	235	240	31.58%	80.29484558	25.35626703	
Peru	11	10	1.32%	40.80554517	0.536915068	
Russia	5	5	0.66%	26.78473822	0.176215383	
World	741	760			67.05813082	

	Usable Ore		Iron Content		Reserves						
Iron Ore	2018	2019	2018	2019	Crude Ore	Iron Content	Weightd Mining	Governance Index	Weighted index	Recycling Score %:	
United States	49,500	48,000	31,300	31,000	3,000	1,000	1.28%	78.99384054	1.01108476	71.8	
Australia	900,000	930,000	557,000	580,000	848,000	823,000	56.12%	91.18427531	51.17077316		
Brazil	460,000	480,000	250,000	260,000	29,000	15,000	11.00%	40.81806405	4.489825412		
Canada	52,400	54,000	31,500	33,000	6,000	2,300	1.41%	91.72706858	1.295139583		
Chile	14,000	14,000	8,940	9,000			0.36%	73.4654541	0.264466114		
China	335,000	350,000	209,000	220,000	20,000	6,900	9.08%	44.06116374	3.998847263		
India	205,000	210,000	126,000	130,000	5,500	3,400	5.34%	47.7993412	2.550481029		
Iran	36,400	38,000	23,900	25,000	2,700	1,500	1.06%	11.8421394	0.125522159		
Kazakhstan	41,900	43,000	11,700	12,000	2,500	900	0.52%	42.04158465	0.216926767		
Mexico	22,300	23,000	14,000	14,000			0.56%	31.93078709	0.178805971		
Peru	14,200	15,000	9,530	10,000			0.40%	40.80554517	0.163216305		
Russia	96,100	99,000	56,700	59,000	25,000	14,000	2.92%	26.78473822	0.782086201		
South Africa	74,300	77,000	47,200	49,000	1,100	690	1.99%	51.35069784	1.020609728		
Sweden	35,800	37,000	22,200	23,000	1,300	600	0.94%	94.7515475	0.894422409		
Ukraine	60,300	62,000	37,700	39,000	96,500	92,300	5.25%	31.89271243	1.674944959		
Other	62,500	62,000	35,800	35,000	18,000	9,500	1.78%				
World	2,460,000	2,500,000	1,470,000	1,529,000	170,000	971,090			69.83715182		

Lead	2018	2019	Reserves	Weightd Mining	Governance Index	Weighted index	Recycling Score %:
United States	280	280	5,000	0.59%	78.99384054	0.466092437	71.7
Australia	432	430	836,000	93.47%	91.18427531	85.2303862	
Bolivia	112	100	1,600	0.19%	24.04151487	0.045672592	
China	2,100	2,100	18,000	2.25%	44.06116374	0.989684857	
India	192	190	2,500	0.30%	47.7993412	0.143687535	
Kazakhstan	86	90	2,000	0.23%	42.04158465	0.09819068	
Mexico	240	240	5,600	0.65%	31.93078709	0.208385442	
Peru	289	290	6,300	0.74%	40.80554517	0.300503478	
Russia	220	220	6,400	0.74%	26.78473822	0.198148277	
Sweden	65	60	1,100	0.13%	94.7515475	0.122825688	
Turkey	76	70	860	0.10%	35.23606904	0.036619744	
Other	468	430	5,000	0.61%			
World	4,560	4,500	890,360	100.00%		87.84019693	

Lithium	2018	2019	Reserves	Weightd Mining	Governance Index	Weighted index	Recycling Score %:
United States			630,000	0.82%	78.99384054	0.649161119	1
Argentina	6,400	6,400	1,700,000	2.23%	42.21330039	0.939612688	
Australia	58,800	42,000	62,800,000	81.97%	91.18427531	74.74612298	
Brazil	300	300	95,000	0.12%	40.81806405	0.050741585	
Canada	2,400	200	370,000	0.48%	91.72706858	0.442947904	
Chile	17,000	18,000	8,600,000	11.24%	73.4654541	8.258637026	
China	7,100	7,500	1,000,000	1.31%	44.06116374	0.579054899	
Namibia	500	0		0.00%	60.10453097	0	
Portugal	800	1,200	60,000	0.08%	81.46109009	0.065030989	
Zimbabwe	1,600	1,600	230,000	0.30%	11.34755913	0.034281493	
Other7	—	0	1,100,000	1.43%			
World	895,000	877,000	76,585,000			85.76559068	

Manganese	2018	2019	Reserves	Weightd Mining	Governance Index	Weighted index	Recycling Score %:
United States	—	0	0	0.00%	78.99384054	0	53.0
Australia	3,480	3,200	12,100,000	94.33%	91.18427531	86.01212072	
Brazil	1,310	1,200	140,000	1.10%	40.81806405	0.449186396	
Burma	207	210		0.00%	8.562470357	0.000140139	
China	1,200	1,300	54,000	0.43%	44	0.189898087	
Cote d'Ivoire	395	400		0.00%	32.82674567	0.001023357	
Gabon	2,330	2,400	61,000	0.49%	26.47298304	0.130807195	
Georgia	200	200		0.00%	60.99433295	0.000950734	
Ghana	1,360	1,400	13,000	0.11%	50.77295049	0.056981567	
India	961	1,000	34,000	0.27%	47.7993412	0.130385546	
Kazakhstan,	140	130	5,000	0.04%	42.04158465	0.01680877	
Malaysia	390	420		0.00%	62.52833112	0.002046754	
Mexico	210	190	5,000	0.04%	31.93078709	0.012915656	
South Africa	5,800	5,500	260,000	2.07%	51.35069784	1.062552434	
Ukraine,	517	540	140,000	1.10%	31.89271243	0.349325992	
Other	397	910	0	0.01%			
World	18,900	19,000	12,812,000	1		88.41514334	

Molybdenum	2018	2019	Reserves	Weightd Mining	Governance Index	Weighted index	Recycling Score %:
United States	41,400	44,000	2,700	15.08%	78.99384054	11.91183641	30.0
Argentinae	600	600	100	0.23%	42.21330039	0.095414847	
Armenia	5,000	5,400	150	1.79%	46.211085	0.828147623	
Canada	4,680	4,700	100	1.55%	91.72706858	1.421698034	
Chile	60,200	54,000	1,400	17.89%	73.4654541	13.14200243	
China	133,000	130,000	8,300	44.66%	44.06116374	19.67645037	
Iran	3,500	3,500	43	1.14%	11.8421394	0.135478361	
Mexico	15,100	16,000	130	5.21%	31.93078709	1.663077938	
Mongolia	1,800	1,800	210	0.65%	47.26727867	0.306778746	
Peru	28,000	28,000	2,900	9.98%	40.80554517	4.071423461	
Russiae	2,800	2,800	1,000	1.23%	26.78473822	0.328654523	
Turkeye	900	900	700	0.52%	35.23606904	0.182043864	
Uzbekistane	200	200	60	0.08%	28.41022205		
World	297,000	290,000	18,000	99.45%		53.7630066	

Neodymium	2018	2019	Reserves	Weightd Mining	Governance Index	Weighted index	Recycling Score %:
United States	—	—	210,000	2%	78.99384054	1.287533201	1
Brazil	59,000	65,000	11,000,000	86%	40.81806405	35.05498085	
Canada	7,700	7,600	1,600,000	12%	91.72706858	11.44514832	
Other	1,460	1,500		0%			
World	68,200	74,000	13,000,000			47.78766237	



Nickel	2018	2019	Reserves	Weightd Mining	Governance Index	Weighted index	Recycling Score %:
United States	17,600	14,000	110,000	0.014%	78.99384054	0.010988697	59.3
Australia	170,000	180,000	820,000,000	92.011%	91.18427531	83.89969724	
Brazil	74,400	67,000	11,000,000	1.242%	40.81806405	0.506773131	
Canada	176,000	180,000	2,600,000	0.312%	91.72706858	0.286070832	
China	110,000	110,000	2,800,000	0.326%	44.06116374	0.143840181	
Cuba	51,000	51,000	5,500,000	0.623%	35.31973966	0.219947986	
Indonesia	606,000	800,000	21,000,000	2.446%	48.91132736	1.196181855	
New Caledonia	216,000	220,000	NA	0.025%	81.86018372	0.020203502	
Philippines	345,000	420,000	4,800,000	0.586%	38.35872523	0.224629058	
Russia	272,000	270,000	6,900,000	0.804%	26.78473822	0.2154457	
Other	366,000	370,000	14,000,000	1.612%			
World	2,400,000	2,700,000	888,710,000	100.002%		86.72377818	

Silver	2018	2019	Reserves	Weightd Mining	Governance Index	Weighted index	Recycling Score %:
United States	934	980	25,000	1.63%	78.99384054	1.290583442	58.8
Argentina	1,020	1,200		0.08%	42.21330039	0.031855488	
Australia	1,220	1,400	1,090,000	68.63%	91.18427531	62.58317805	
Bolivia	1,190	1,200	22,000	1.46%	24.04151487	0.350754723	
Chile	1,370	1,300	26,000	1.72%	73.4654541	1.261245203	
China	3,570	3,600	41,000	2.80%	44.06116374	1.235789598	
Mexico	6,120	6,300	37,000	2.72%	31.93078709	0.869463256	
Peru	4,160	3,800	120,000	7.79%	40.80554517	3.176826832	
Poland	1,470	1,700	100,000	6.40%	66.6814785	4.264615555	
Russia	2,100	2,100	45,000	2.96%	26.78473822	0.793344886	
Other	3,730	3,600	57,000	3.81%			
World	26,900	27,000	1,563,000			75.85765703	

Titanium	Sponge Production		Capacity		Weightd Mining	Governance Index	Weighted index	Recycling Score %:
	2018	2019	Sponge	Pigment				
United States			13,100	1,370,000	17.36%	78.99384054	13.71275568	91.0
Australia				260,000	3.26%	91.18427531	2.97557723	
Canada				104,000	1.31%	91.72706858	1.197315988	
China	75,000	84,000	117,000	3,250,000	42.26%	44.06116374	18.61988558	
Germany				472,000	5.92%	89.43319956	5.298082233	
India	250	250	500	108,000	1.36%	47.7993412	0.650922939	
Japan	49,000	54,000	68,800	314,000	4.80%	88.73260625	4.263174355	
Kazakhstan	16,000	20,000	31,000	1,000	0.40%	42.04158465	0.168852301	
Mexico				300,000	3.77%	31.93078709	1.202288814	
Russia	44,000	44,000	46,500	55,000	1.27%	26.78473822	0.341217563	
Saudi Arabia			15,600	210,000	2.83%	47.99729188	1.35904475	
Ukraine	8,000	9,000	12,000	120,000	1.66%	31.89271243	0.528376284	
United Kingdom				315,000	3.95%	85.93093872	3.397332375	
Other				784,000	9.84%			
World	6,192,000	6,210,000	305,000	7,660,000			53.71482608	

Vanadium	2018	2019	Reserves	Weightd Mining	Governance Index	Weighted index	Recycling Score %:
United States	—	470	45	0.28%	78.99384054	0.219149556	1
Australia	—	—	94,000	50.64%	91.18427531	46.17298397	
Brazil	5,500	7,000	120	3.84%	40.81806405	1.565570157	
China	40,000	40,000	9,500	26.67%	44.06116374	11.74901072	
Russia	18,000	18,000	5,000	12.39%	26.78473822	3.318603599	
South Africa	7,700	8,000	3,500	6.19%	51.35069784	3.181151319	
World	71,200	73,000	22,000			66.20646932	

Zinc	2018	2019	Reserves	Weightd Mining	Governance Index	Weighted index	Recycling Score %:
United States	824	780	11,000	0.86%	78.99384054	0.682795202	41.5
Australia	1,110	1,300	1,168,000	85.80%	91.18427531	78.23441547	
Bolivia	480	460	4,800	0.39%	24.04151487	0.092789645	
Canada	287	300	2,200	0.18%	91.72706858	0.168263324	
China	4,170	4,300	44,000	3.54%	44.06116374	1.561546912	
India	750	800	7,500	0.61%	47.7993412	0.291106528	
Kazakhstan	304	290	12,000	0.90%	42.04158465	0.379125418	
Mexico	691	690	22,000	1.66%	31.93078709	0.531613574	
Peru	1,470	1,400	19,000	1.50%	40.80554517	0.610803186	
Russia	300	300	22,000	1.64%	26.78473822	0.43827249	
Sweden	234	230	3,600	0.28%	94.7515475	0.266279067	
Other	1,840	1,900	34,000	2.63%			
World	12,500	13,000	1,339,100			83.25701081	



User: Keisi Kapaj  
Project: Thesis - Part 2

#### 1 . ivregress 2sls Lifeexpectancyatbirth NationalIncome GDPperCapita (kwhpercapita = Residual)

Instrumental variables (2SLS) regression

Number of obs	=	<b>146</b>
Wald chi2(3)	=	<b>464.23</b>
Prob > chi2	=	<b>0.0000</b>
R-squared	=	<b>0.7634</b>
Root MSE	=	<b>.02419</b>

Lifeexpectan~h	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
kwhpercapita	<b>.0343612</b>	<b>.0310016</b>	<b>1.11</b>	<b>0.268</b>	<b>-.0264007</b>	<b>.0951232</b>
NationalIncome	<b>.1304962</b>	<b>.0260942</b>	<b>5.00</b>	<b>0.000</b>	<b>.0793526</b>	<b>.1816399</b>
GDPperCapita	<b>-.0915458</b>	<b>.0386058</b>	<b>-2.37</b>	<b>0.018</b>	<b>-.1672117</b>	<b>-.0158799</b>
_cons	<b>1.583083</b>	<b>.0178272</b>	<b>88.80</b>	<b>0.000</b>	<b>1.548143</b>	<b>1.618024</b>

Instrumented: kwhpercapita  
Instruments: NationalIncome GDPperCapita Residual

#### 2 . estat endog

Tests of endogeneity  
Ho: variables are exogenous

Durbin (score) chi2(1)	=	<b>.127513</b>	(p = <b>0.7210</b> )
Wu-Hausman F(1,141)	=	<b>.123254</b>	(p = <b>0.7261</b> )

#### 3 . estat firststage

First-stage regression summary statistics

Variable	R-sq.	Adjusted R-sq.	Partial R-sq.	F(1,142)	Prob > F
kwhpercapita	<b>0.8417</b>	<b>0.8384</b>	<b>0.0579</b>	<b>8.73301</b>	<b>0.0037</b>

Minimum eigenvalue statistic = **8.73301**

Critical Values	# of endogenous regressors:	<b>1</b>
Ho: Instruments are weak	# of excluded instruments:	<b>1</b>

2SLS relative bias	5%	10%	20%	30%
	(not available)			
2SLS Size of nominal 5% Wald test	10%	15%	20%	25%
LIML Size of nominal 5% Wald test	<b>16.38</b>	<b>8.96</b>	<b>6.66</b>	<b>5.53</b>

#### 4 . ivregress 2sls Expectedyearsofschooling NationalIncome GDPperCapita (kwhpercapita = Residual)

Instrumental variables (2SLS) regression

Number of obs	=	<b>145</b>
Wald chi2(3)	=	<b>312.94</b>
Prob > chi2	=	<b>0.0000</b>
R-squared	=	<b>0.6868</b>
Root MSE	=	<b>.05477</b>

Expectedyear~g	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
kwhpercapita	<b>.0531675</b>	<b>.0722325</b>	<b>0.74</b>	<b>0.462</b>	<b>-.0884055</b>	<b>.1947406</b>
NationalIncome	<b>.1545424</b>	<b>.0610559</b>	<b>2.53</b>	<b>0.011</b>	<b>.034875</b>	<b>.2742099</b>
GDPperCapita	<b>-.0704628</b>	<b>.0960966</b>	<b>-0.73</b>	<b>0.463</b>	<b>-.2588086</b>	<b>.1178831</b>
_cons	<b>.6137938</b>	<b>.0439845</b>	<b>13.95</b>	<b>0.000</b>	<b>.5275858</b>	<b>.7000017</b>

Instrumented: kwhpercapita  
Instruments: NationalIncome GDPperCapita Residual

5 . estat endog

Tests of endogeneity  
Ho: variables are exogenous

Durbin (score) chi2(1) = .024806 (p = 0.8749)  
Wu-Hausman F(1,140) = .023955 (p = 0.8772)

6 . estat firststage

First-stage regression summary statistics

Variable	R-sq.	Adjusted R-sq.	Partial R-sq.	F(1,141)	Prob > F
kwhpercapita	0.8399	0.8365	0.0566	8.45928	0.0042

Minimum eigenvalue statistic = 8.45928

Critical Values # of endogenous regressors: 1  
Ho: Instruments are weak # of excluded instruments: 1

	5%	10%	20%	30%
2SLS relative bias	(not available)			
2SLS Size of nominal 5% Wald test	16.38	8.96	6.66	5.53
LIML Size of nominal 5% Wald test	16.38	8.96	6.66	5.53

7 . ivregress 2sls Meanyearsofschooling NationalIncome GDPperCapita (kwhpercapita = Residual)

Instrumental variables (2SLS) regression

Number of obs	=	145
Wald chi2(3)	=	350.14
Prob > chi2	=	0.0000
R-squared	=	0.7240
Root MSE	=	.10455

Meanyearsofs~g	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
kwhpercapita	.1488715	.1378936	1.08	0.280	-.1213951	.4191381
NationalIncome	.238623	.1165573	2.05	0.041	.0101748	.4670712
GDPperCapita	-.1098591	.1834509	-0.60	0.549	-.4694162	.2496981
_cons	-.142776	.0839675	-1.70	0.089	-.3073493	.0217972

Instrumented: kwhpercapita  
Instruments: NationalIncome GDPperCapita Residual

8 . estat endog

Tests of endogeneity  
Ho: variables are exogenous

Durbin (score) chi2(1) = .086392 (p = 0.7688)  
Wu-Hausman F(1,140) = .083463 (p = 0.7731)

9 . estat firststage

First-stage regression summary statistics

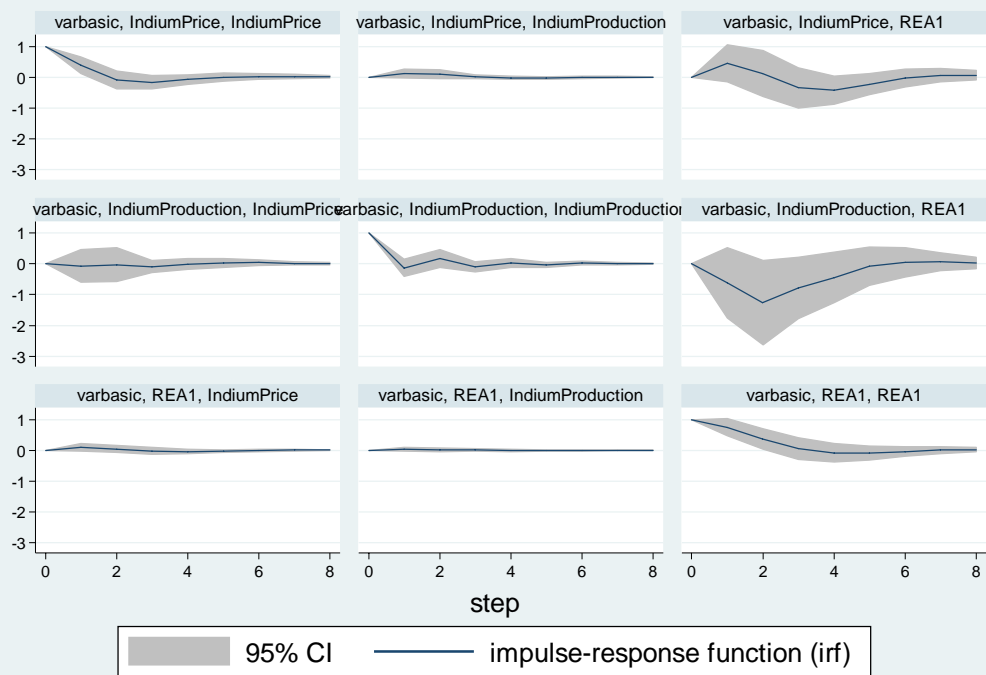
Variable	R-sq.	Adjusted R-sq.	Partial R-sq.	F(1,141)	Prob > F
kwhpercapita	0.8399	0.8365	0.0566	8.45928	0.0042

Minimum eigenvalue statistic = 8.45928

Critical Values	# of endogenous regressors:	<b>1</b>
Ho: Instruments are weak	# of excluded instruments:	<b>1</b>

	5%	10%	20%	30%
2SLS relative bias	(not available)			
	10%	15%	20%	25%
2SLS Size of nominal 5% Wald test	<b>16.38</b>	<b>8.96</b>	<b>6.66</b>	<b>5.53</b>
LIML Size of nominal 5% Wald test	<b>16.38</b>	<b>8.96</b>	<b>6.66</b>	<b>5.53</b>





Graphs by irfname, impulse variable, and response variable

(R)

Statistics/Data Analysis

User: Keisi Kapaj  
Project: Thesis - Part 3

1 . varbasic IndiumPrice IndiumProduction REA1, lags(1/2) step(8) irf

Vector autoregression

Sample: 1970 - 2014	Number of obs	=	45
Log likelihood = 53.95139	AIC	=	-1.464506
FPE = .0000468	HQIC	=	-1.150203
Det(Sigma_ml) = .0000182	SBIC	=	-.6213969

Equation	Parms	RMSE	R-sq	chi2	P>chi2
IndiumPrice	7	.17337	0.1870	10.35263	0.1106
IndiumProduction	7	.093334	0.1307	6.768214	0.3428
REA1	7	.376775	0.4694	39.81101	0.0000

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
<b>IndiumPrice</b>						
IndiumPrice						
L1.	.3913282	.1396921	2.80	0.005	.1175368	.6651196
L2.	-.2774656	.1426399	-1.95	0.052	-.5570345	.0021034
IndiumProduction						
L1.	-.0736547	.2702025	-0.27	0.785	-.6032418	.4559325
L2.	.0490905	.2710923	0.18	0.856	-.4822407	.5804217
REA1						
L1.	.098126	.0662785	1.48	0.139	-.0317774	.2280295
L2.	-.0645198	.0669777	-0.96	0.335	-.1957937	.0667541
_cons	.0056514	.0260893	0.22	0.829	-.0454827	.0567855
<b>IndiumProduction</b>						
IndiumPrice						
L1.	.1121667	.075203	1.49	0.136	-.0352285	.2595618
L2.	.0591405	.0767899	0.77	0.441	-.091365	.209646
IndiumProduction						
L1.	-.1347798	.1454631	-0.93	0.354	-.4198822	.1503225
L2.	.1867694	.1459421	1.28	0.201	-.0992719	.4728107
REA1						
L1.	.0360503	.0356809	1.01	0.312	-.033883	.1059836
L2.	-.0190595	.0360573	-0.53	0.597	-.0897306	.0516116
_cons	.0226652	.0140451	1.61	0.107	-.0048628	.0501932
<b>REA1</b>						
IndiumPrice						
L1.	.4530836	.3035842	1.49	0.136	-.1419304	1.048098
L2.	-.3320257	.3099904	-1.07	0.284	-.9395958	.2755443
IndiumProduction						
L1.	-.6174783	.5872144	-1.05	0.293	-1.768397	.5334408
L2.	-.8471059	.5891482	-1.44	0.150	-2.001815	.3076034
REA1						
L1.	.7469936	.1440389	5.19	0.000	.4646825	1.029305
L2.	-.2163047	.1455585	-1.49	0.137	-.5015941	.0689848
_cons	.0603782	.0566983	1.06	0.287	-.0507484	.1715049

2 . varstable

Eigenvalue stability condition

Eigenvalue	Modulus
.4132601 + .482703i	.635442
.4132601 - .482703i	.635442
-.534862	.534862
.4829338	.482934
.114475 + .3727593i	.389941
.114475 - .3727593i	.389941

All the eigenvalues lie inside the unit circle.  
VAR satisfies stability condition.

3 . predict error44, resid  
(10 missing values generated)

4 . summarize error44

Variable	Obs	Mean	Std. Dev.	Min	Max
error44	45	1.24e-10	.1611165	-.265579	.4872707

5 . tsline error44, yline(1.24e-10 )

6 . varlmar

Lagrange-multiplier test

lag	chi2	df	Prob > chi2
1	6.2785	9	0.71175
2	15.2943	9	0.08316

H0: no autocorrelation at lag order

7 . vargranger

Granger causality Wald tests

Equation	Excluded	chi2	df	Prob > chi2
IndiumPrice	IndiumProduction	.12041	2	0.942
IndiumPrice	REAL	2.1927	2	0.334
IndiumPrice	ALL	2.3191	4	0.677
IndiumProduction	IndiumPrice	3.7654	2	0.152
IndiumProduction	REAL	1.0428	2	0.594
IndiumProduction	ALL	4.5874	4	0.332
REAL	IndiumPrice	2.6866	2	0.261
REAL	IndiumProduction	2.8563	2	0.240
REAL	ALL	7.4342	4	0.115

8 . irf table fevd, impulse( IndiumPrice IndiumProduction REAL ) response( IndiumPrice IndiumProduction REAL )  
> i

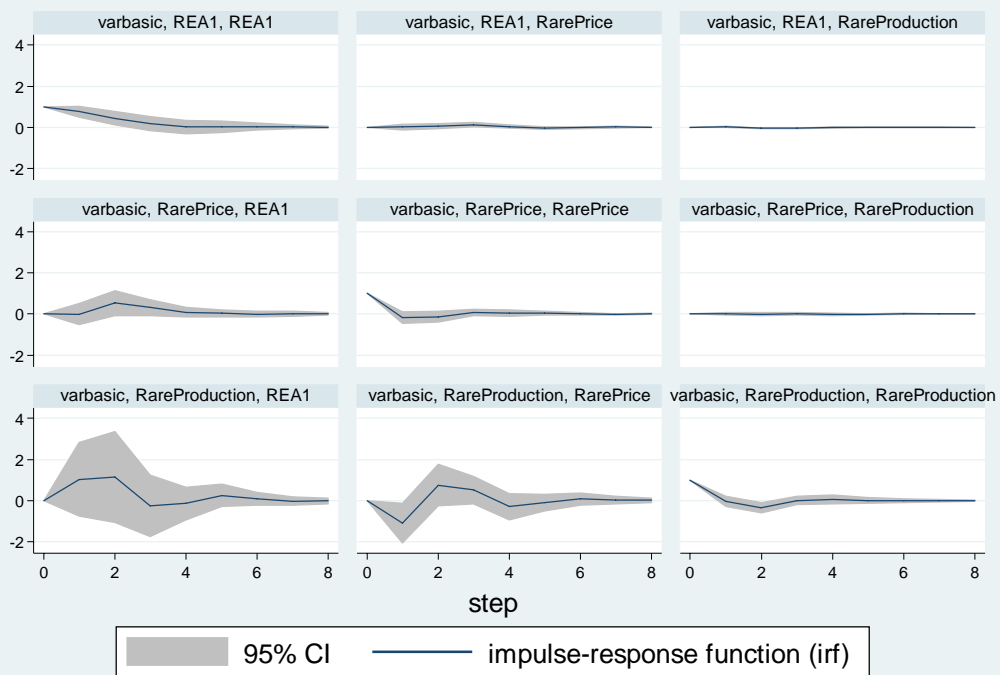
Results from varbasic

step	(1) fevd	(2) fevd	(3) fevd	(4) fevd	(5) fevd	(6) fevd	(7) fevd	(8) fevd
0	0	0	0	0	0	0	0	0
1	1	.050863	.137219	0	.949137	.00381	0	0
2	.964912	.063184	.098485	.000573	.919466	.009598	.034515	.01735
3	.958612	.098733	.099624	.000647	.882673	.054475	.040741	.018594
4	.955724	.097995	.122367	.002955	.88098	.071187	.041321	.021025
5	.950807	.098462	.138138	.003023	.880351	.075837	.04617	.021187
6	.949468	.099376	.139947	.003096	.878984	.075597	.047437	.02164
7	.949218	.099315	.139868	.003398	.878584	.075577	.047384	.022101
8	.948997	.099365	.140351	.003459	.878515	.075696	.047544	.02212



step	(9) fevd
0	0
1	.858971
2	.891918
3	.845901
4	.806447
5	.786025
6	.784456
7	.784554
8	.783953

(1) irfname = varbasic, impulse = IndiumPrice, and response = IndiumPrice  
(2) irfname = varbasic, impulse = IndiumPrice, and response = IndiumProduction  
(3) irfname = varbasic, impulse = IndiumPrice, and response = REAL  
(4) irfname = varbasic, impulse = IndiumProduction, and response = IndiumPrice  
(5) irfname = varbasic, impulse = IndiumProduction, and response = IndiumProduction  
(6) irfname = varbasic, impulse = IndiumProduction, and response = REAL  
(7) irfname = varbasic, impulse = REAL, and response = IndiumPrice  
(8) irfname = varbasic, impulse = REAL, and response = IndiumProduction  
(9) irfname = varbasic, impulse = REAL, and response = REAL



Graphs by irfname, impulse variable, and response variable

(R)

Statistics/Data Analysis

User: Keisi Kapaj  
Project: Thesis - Part 3

1 . varbasic RarePrice RareProduction REA1, lags(1/2) step(8) irf

Vector autoregression

Sample: 1970 - 2019	Number of obs	=	50
Log likelihood = 77.07347	AIC	=	-2.242939
FPE = .0000214	HQIC	=	-1.937133
Det(Sigma_ml) = 9.20e-06	SBIC	=	-1.439889

Equation	Parms	RMSE	R-sq	chi2	P>chi2
RarePrice	7	.203998	0.2054	12.92648	0.0442
RareProduction	7	.049966	0.3342	25.10288	0.0003
REA1	7	.376431	0.4930	48.61953	0.0000

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
RarePrice						
RarePrice						
L1.	-.1851074	.1399667	-1.32	0.186	-.459437	.0892222
L2.	-.151043	.1327264	-1.14	0.255	-.4111819	.1090958
RareProduction						
L1.	-1.099868	.492721	-2.23	0.026	-2.065583	-.1341522
L2.	.4913093	.5278916	0.93	0.352	-.5433391	1.525958
REAL						
L1.	.0193838	.0732901	0.26	0.791	-.1242622	.1630298
L2.	.0834829	.0747177	1.12	0.264	-.0629611	.2299269
_cons	.0456845	.0319127	1.43	0.152	-.0168634	.1082323
RareProduction						
RarePrice						
L1.	.0143246	.0342827	0.42	0.676	-.0528683	.0815175
L2.	-.0037957	.0325093	-0.12	0.907	-.0675128	.0599214
RareProduction						
L1.	-.0364317	.1206846	-0.30	0.763	-.2729692	.2001057
L2.	-.37659	.1292991	-2.91	0.004	-.6300115	-.1231684
REAL						
L1.	.0303275	.0179513	1.69	0.091	-.0048564	.0655114
L2.	-.0732706	.018301	-4.00	0.000	-.1091399	-.0374014
_cons	.0295097	.0078165	3.78	0.000	.0141896	.0448298
REAL						
RarePrice						
L1.	-.0059702	.2582768	-0.02	0.982	-.5121833	.5002429
L2.	.5199486	.2449164	2.12	0.034	.0399212	.999976
RareProduction						
L1.	1.030903	.909205	1.13	0.257	-.7511057	2.812913
L2.	.391413	.9741043	0.40	0.688	-1.517796	2.300622
REAL						
L1.	.7656831	.1352403	5.66	0.000	.500617	1.030749
L2.	-.1720763	.1378746	-1.25	0.212	-.4423055	.0981528
_cons	-.0424623	.0588877	-0.72	0.471	-.1578802	.0729555

2 . varstable

Eigenvalue stability condition

Eigenvalue	Modulus
.1482841 + .6060266i	.623904
.1482841 - .6060266i	.623904
-.2557742 + .4883768i	.551301
-.2557742 - .4883768i	.551301
.4098917	.409892
.3492326	.349233

All the eigenvalues lie inside the unit circle.  
VAR satisfies stability condition.

3 . predict error33, resid  
(5 missing values generated)

4 . summarize error33

Variable	Obs	Mean	Std. Dev.	Min	Max
error33	50	2.14e-10	.1911003	-.39514	.5868568

5 . tsline error33, yline(2.14e-10)

6 . varlmar

Lagrange-multiplier test

lag	chi2	df	Prob > chi2
1	8.8846	9	0.44800
2	13.9454	9	0.12428

H0: no autocorrelation at lag order

7 . vargranger

Granger causality Wald tests

Equation	Excluded	chi2	df	Prob > chi2
RarePrice	RareProduction	6.1054	2	0.047
RarePrice	REAL	2.9679	2	0.227
RarePrice	ALL	10.471	4	0.033
RareProduction	RarePrice	.21198	2	0.899
RareProduction	REAL	17.514	2	0.000
RareProduction	ALL	17.94	4	0.001
REAL	RarePrice	4.6696	2	0.097
REAL	RareProduction	1.3997	2	0.497
REAL	ALL	6.4613	4	0.167

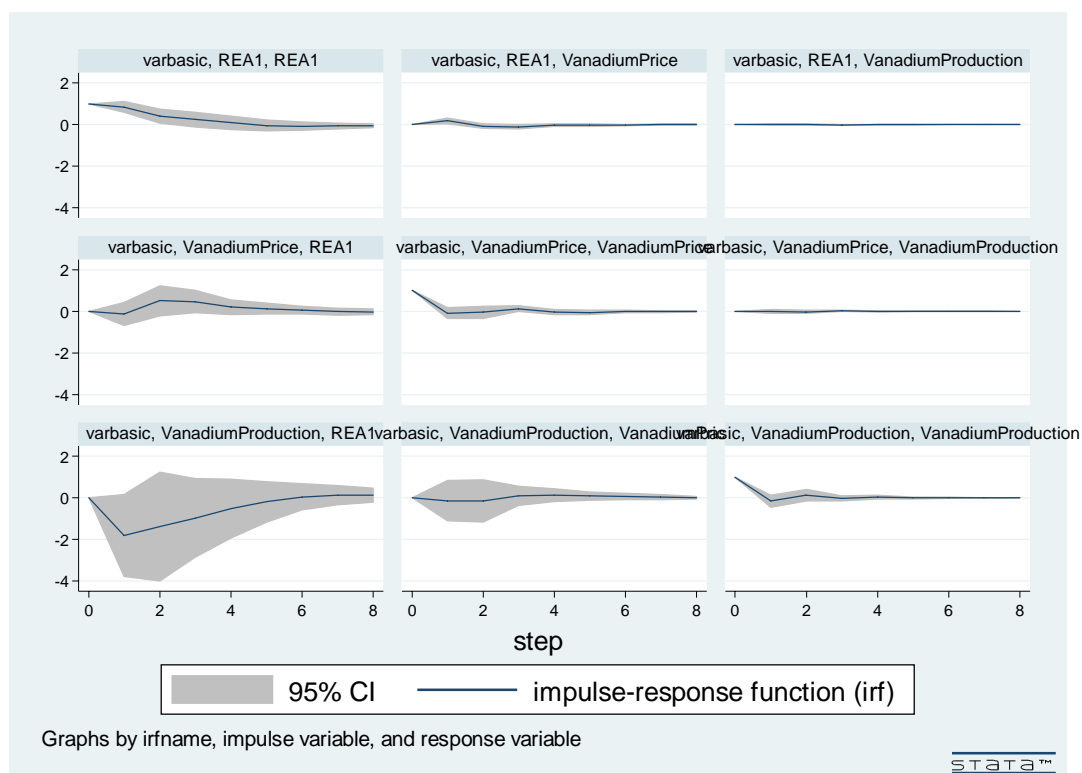
8 . irf table fevd, impulse( RarePrice RareProduction REAL ) response( RarePrice RareProduction REAL ) noc

Results from varbasic

step	(1) fevd	(2) fevd	(3) fevd	(4) fevd	(5) fevd	(6) fevd	(7) fevd	(8) fevd
0	0	0	0	0	0	0	0	0
1	1	.016332	.001477	0	.983668	5.4e-06	0	0
2	.933597	.020313	.000968	.065236	.930449	.01119	.001167	.049238
3	.897754	.016147	.043592	.090148	.83459	.021275	.012098	.149262
4	.846058	.014992	.058508	.096932	.750021	.021173	.05701	.234987
5	.839427	.023184	.059636	.100051	.74354	.021328	.060522	.233276
6	.837002	.028599	.059657	.100048	.738	.021857	.062951	.233401
7	.836652	.028935	.059602	.100251	.737742	.021904	.063097	.233323
8	.836344	.029349	.059614	.100202	.737171	.021906	.063454	.23348

step	(9) fevd
0	0
1	.998518
2	.987842
3	.935133
4	.920318
5	.919036
6	.918486
7	.918494
8	.918479

(1) irfname = varbasic, impulse = RarePrice, and response = RarePrice  
 (2) irfname = varbasic, impulse = RarePrice, and response = RareProduction  
 (3) irfname = varbasic, impulse = RarePrice, and response = REAL  
 (4) irfname = varbasic, impulse = RareProduction, and response = RarePrice  
 (5) irfname = varbasic, impulse = RareProduction, and response = RareProduction  
 (6) irfname = varbasic, impulse = RareProduction, and response = REAL  
 (7) irfname = varbasic, impulse = REAL, and response = RarePrice  
 (8) irfname = varbasic, impulse = REAL, and response = RareProduction  
 (9) irfname = varbasic, impulse = REAL, and response = REAL



(R)

Statistics/Data Analysis

User: Keisi Kapaj  
Project: Thesis - Part 3

1 . varbasic VanadiumPrice VanadiumProduction REA1, lags(1/2) step(8) irf

Vector autoregression

Sample: 1970 - 2019	Number of obs	=	50
Log likelihood = 84.04813	AIC	=	-2.521925
FPE = .0000162	HQIC	=	-2.216119
Det(Sigma_ml) = 6.96e-06	SBIC	=	-1.718875

Equation	Parms	RMSE	R-sq	chi2	P>chi2
VanadiumPrice	7	.180665	0.1897	11.70639	0.0688
VanadiumProduc~n	7	.053999	0.0538	2.841185	0.8285
REA1	7	.368186	0.5150	53.08636	0.0000

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
<b>VanadiumPrice</b>						
VanadiumPrice						
L1.	-.0870717	.1370181	-0.64	0.525	-.3556223	.1814789
L2.	-.0395604	.1422833	-0.28	0.781	-.3184305	.2393097
<b>VanadiumProduction</b>						
L1.	-.1397534	.4960095	-0.28	0.778	-1.111914	.8324073
L2.	.1537236	.5019498	0.31	0.759	-.83008	1.137527
<b>REA1</b>						
L1.	.1818439	.0666095	2.73	0.006	.0512917	.312396
L2.	-.2069445	.0668446	-3.10	0.002	-.3379576	-.0759314
_cons	.0205741	.0270157	0.76	0.446	-.0323757	.0735239
<b>VanadiumProduction</b>						
VanadiumPrice						
L1.	-.0109584	.0409533	-0.27	0.789	-.0912255	.0693086
L2.	-.0274083	.042527	-0.64	0.519	-.1107598	.0559432
<b>VanadiumProduction</b>						
L1.	-.1524282	.1482522	-1.03	0.304	-.4429972	.1381409
L2.	.1303844	.1500277	0.87	0.385	-.1636645	.4244334
<b>REA1</b>						
L1.	.0144648	.0199089	0.73	0.468	-.0245559	.0534855
L2.	-.0147656	.0199792	-0.74	0.460	-.0539241	.0243928
_cons	.0193673	.0080747	2.40	0.016	.0035411	.0351934
<b>REA1</b>						
VanadiumPrice						
L1.	-.1267078	.2792353	-0.45	0.650	-.673999	.4205834
L2.	.5795115	.2899654	2.00	0.046	.0111897	1.147833
<b>VanadiumProduction</b>						
L1.	-1.808436	1.01084	-1.79	0.074	-3.789646	.1727734
L2.	-.1401569	1.022946	-0.14	0.891	-2.145094	1.86478
<b>REA1</b>						
L1.	.8502516	.1357464	6.26	0.000	.5841936	1.11631
L2.	-.2675444	.1362256	-1.96	0.050	-.5345417	-.000547
_cons	.0283389	.0550565	0.51	0.607	-.0795698	.1362476

2 . varstable

Eigenvalue stability condition

Eigenvalue	Modulus
.5912314 + .3261711i	.675235
.5912314 - .3261711i	.675235
-.197975 + .5003265i	.538071
-.197975 - .5003265i	.538071
-.4236271	.423627
.247866	.247866

All the eigenvalues lie inside the unit circle.  
VAR satisfies stability condition.

3 . predict error11, resid  
(5 missing values generated)

4 . summarize error11

Variable	Obs	Mean	Std. Dev.	Min	Max
error11	50	-4.42e-10	.169243	-.3790916	.3782963

5 . tsline error11, yline(-4.42e-10 )

6 . varlmar

Lagrange-multiplier test

lag	chi2	df	Prob > chi2
1	17.8585	9	0.03685
2	12.5871	9	0.18220

H0: no autocorrelation at lag order

7 . vargranger

Granger causality Wald tests

Equation	Excluded	chi2	df	Prob > chi2
VanadiumPrice	VanadiumProduct~n	.20486	2	0.903
VanadiumPrice	REAL	10.391	2	0.006
VanadiumPrice	ALL	10.609	4	0.031
VanadiumProduct~n	VanadiumPrice	.46706	2	0.792
VanadiumProduct~n	REAL	.64515	2	0.724
VanadiumProduct~n	ALL	1.2385	4	0.872
REAL	VanadiumPrice	4.3318	2	0.115
REAL	VanadiumProduct~n	3.2209	2	0.200
REAL	ALL	9.0186	4	0.061

8 . irf table fevd, impulse( VanadiumPrice VanadiumProduction REAL ) response( VanadiumPrice VanadiumProduction ) nci  
> 1 ) nci

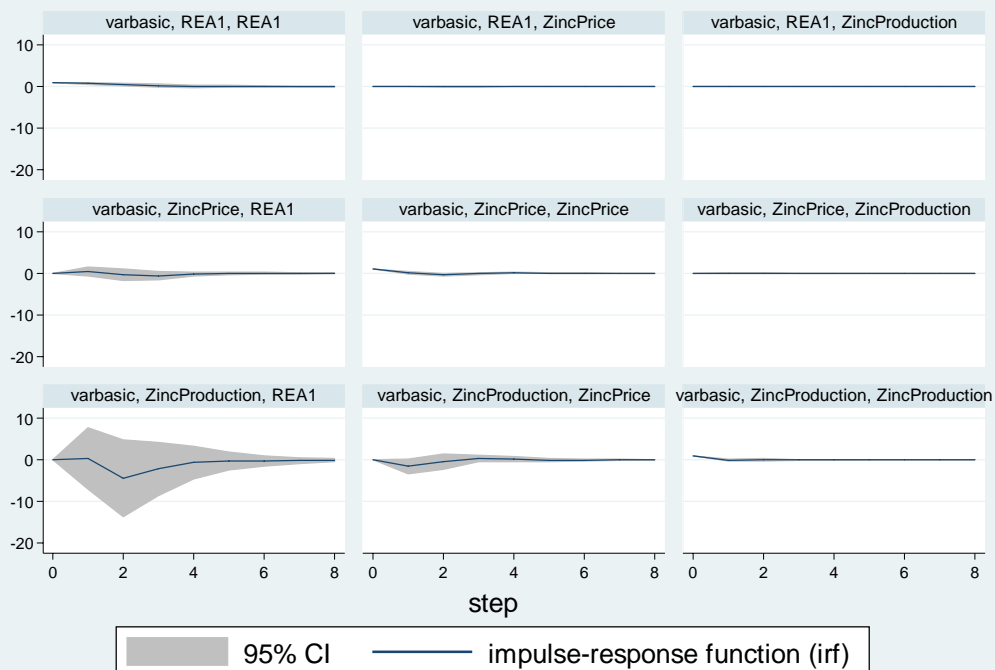
Results from varbasic

step	(1) fevd	(2) fevd	(3) fevd	(4) fevd	(5) fevd	(6) fevd	(7) fevd	(8) fevd
0	0	0	0	0	0	0	0	0
1	1	.114571	.005648	0	.885429	.036698	0	0
2	.883537	.117424	.007848	.000882	.873544	.025893	.11558	.009032
3	.865914	.117786	.029579	.00473	.871443	.029679	.129356	.010771
4	.839299	.118401	.0477	.004621	.868185	.031918	.15608	.013414
5	.836729	.11839	.050115	.005145	.868176	.033156	.158126	.013433
6	.834693	.118618	.05115	.005227	.867944	.033671	.16008	.013439
7	.833305	.118621	.051249	.005294	.867924	.033602	.161401	.013456
8	.833247	.118618	.051196	.005378	.867905	.033546	.161375	.013477



step	(9) fevd
0	0
1	.957654
2	.96626
3	.940742
4	.920382
5	.916729
6	.91518
7	.915149
8	.915258

```
(1) irfname = varbasic, impulse = VanadiumPrice, and response = VanadiumPrice
(2) irfname = varbasic, impulse = VanadiumPrice, and response = VanadiumProduction
(3) irfname = varbasic, impulse = VanadiumPrice, and response = REAL
(4) irfname = varbasic, impulse = VanadiumProduction, and response = VanadiumPrice
(5) irfname = varbasic, impulse = VanadiumProduction, and response = VanadiumProduction
(6) irfname = varbasic, impulse = VanadiumProduction, and response = REAL
(7) irfname = varbasic, impulse = REAL, and response = VanadiumPrice
(8) irfname = varbasic, impulse = REAL, and response = VanadiumProduction
(9) irfname = varbasic, impulse = REAL, and response = REAL
```



Graphs by irfname, impulse variable, and response variable

(R)

Statistics/Data Analysis

User: Keisi Kapaj  
Project: Thesis - Part 3

1 . varbasic ZincPrice ZincProduction REAL, lags(1/2) step(8) irf

Vector autoregression

Sample: 1970 - 2019	Number of obs	=	50
Log likelihood = 178.3668	AIC	=	-6.29467
FPE = 3.73e-07	HQIC	=	-5.988864
Det(Sigma_ml) = 1.60e-07	SBIC	=	-5.491621

Equation	Parms	RMSE	R-sq	chi2	P>chi2
ZincPrice	7	.095864	0.1594	9.482004	0.1482
ZincProduction	7	.014927	0.0748	4.039964	0.6713
REAL	7	.386845	0.4646	43.38163	0.0000

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
ZincPrice						
ZincPrice						
L1.	.0972489	.1440814	0.67	0.500	-.1851454	.3796433
L2.	-.2981991	.1382792	-2.16	0.031	-.5692214	-.0271769
ZincProduction						
L1.	-1.562741	.931103	-1.68	0.093	-3.387669	.2621877
L2.	-.3825191	.9554966	-0.40	0.689	-2.255258	1.49022
REAL						
L1.	.0395836	.0373292	1.06	0.289	-.0335803	.1127474
L2.	-.0053895	.0354144	-0.15	0.879	-.0748004	.0640214
_cons	.0381301	.0178394	2.14	0.033	.0031656	.0730947
ZincProduction						
ZincPrice						
L1.	.0209841	.0224349	0.94	0.350	-.0229875	.0649557
L2.	.000239	.0215315	0.01	0.991	-.0419619	.0424399
ZincProduction						
L1.	-.0535437	.1449821	-0.37	0.712	-.3377033	.2306159
L2.	.0025925	.1487804	0.02	0.986	-.2890117	.2941967
REAL						
L1.	.0068224	.0058125	1.17	0.241	-.00457	.0182147
L2.	-.0011793	.0055144	-0.21	0.831	-.0119873	.0096287
_cons	.0074774	.0027778	2.69	0.007	.002033	.0129217
REAL						
ZincPrice						
L1.	.4170945	.5814159	0.72	0.473	-.7224597	1.556649
L2.	-.7512431	.5580022	-1.35	0.178	-1.844907	.3424212
ZincProduction						
L1.	.297657	3.757308	0.08	0.937	-7.066531	7.661845
L2.	-4.039985	3.855744	-1.05	0.295	-11.5971	3.517134
REAL						
L1.	.7436412	.1506355	4.94	0.000	.448401	1.038881
L2.	-.0920177	.1429087	-0.64	0.520	-.3721137	.1880782
_cons	.0369365	.0719877	0.51	0.608	-.1041569	.1780299

2 . varstable

Eigenvalue stability condition

Eigenvalue	Modulus
.08931262 + .5428354i	.550134
.08931262 - .5428354i	.550134
.4634904	.46349
-.2280627	.228063
.1866468 + .02782901i	.18871
.1866468 - .02782901i	.18871

All the eigenvalues lie inside the unit circle.  
VAR satisfies stability condition.

3 . predict error22, resid  
(5 missing values generated)

4 . summarize error22

Variable	Obs	Mean	Std. Dev.	Min	Max
error22	50	-4.76e-10	.0898036	-.1800298	.3649224

5 . stline error22, yline(-4.76e-10)  
**command stline is unrecognized**  
r(199);

6 . tsline error22, yline(-4.76e-10)

7 . varlmar

Lagrange-multiplier test

lag	chi2	df	Prob > chi2
1	7.2530	9	0.61079
2	11.0325	9	0.27349

H0: no autocorrelation at lag order

8 . vargranger

Granger causality Wald tests

Equation	Excluded	chi2	df	Prob > chi2
ZincPrice	ZincProduction	2.8815	2	0.237
ZincPrice	REAL	1.4225	2	0.491
ZincPrice	ALL	3.2255	4	0.521
ZincProduction	ZincPrice	.88025	2	0.644
ZincProduction	REAL	1.6793	2	0.432
ZincProduction	ALL	4.0292	4	0.402
REAL	ZincPrice	2.2084	2	0.331
REAL	ZincProduction	1.1274	2	0.569
REAL	ALL	3.4625	4	0.484

9 . irf table fevd, impulse( ZincPrice ZincProduction REAL ) response( ZincPrice ZincProduction REAL ) noc

Results from varbasic

step	(1) fevd	(2) fevd	(3) fevd	(4) fevd	(5) fevd	(6) fevd	(7) fevd	(8) fevd
0	0	0	0	0	0	0	0	0
1	1	.008957	.064699	0	.991043	.10716	0	0
2	.947178	.040667	.093061	.033314	.934912	.107048	.019508	.024421
3	.945834	.043147	.084037	.032647	.922907	.095743	.021519	.033946
4	.944712	.044545	.085348	.033748	.918338	.092943	.02154	.037117
5	.944377	.045198	.085496	.033961	.917354	.092492	.021662	.037448
6	.944371	.045237	.085521	.033979	.917277	.092404	.02165	.037486
7	.944289	.045306	.085548	.034025	.91719	.092379	.021687	.037504
8	.944298	.045306	.085549	.03402	.91718	.092375	.021683	.037514

step	(9) fevd
0	0
1	.828141
2	.799891
3	.82022
4	.821709
5	.822012
6	.822075
7	.822073
8	.822076

(1) irfname = varbasic, impulse = ZincPrice, and response = ZincPrice  
(2) irfname = varbasic, impulse = ZincPrice, and response = ZincProduction  
(3) irfname = varbasic, impulse = ZincPrice, and response = REAL  
(4) irfname = varbasic, impulse = ZincProduction, and response = ZincPrice  
(5) irfname = varbasic, impulse = ZincProduction, and response = ZincProduction  
(6) irfname = varbasic, impulse = ZincProduction, and response = REAL  
(7) irfname = varbasic, impulse = REAL, and response = ZincPrice  
(8) irfname = varbasic, impulse = REAL, and response = ZincProduction  
(9) irfname = varbasic, impulse = REAL, and response = REAL