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## **Semiconductor Supply Chain Disruptions – Its Effect on Automotive Industry: China and U.S.**

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### **ABSTRACT**

This paper analyses and defines the causes of supply chain disruptions in the semiconductor industry and explains how the automotive industry was therefore affected, including the industries' challenges and opportunities. In the research, China and the U.S. were primarily analysed. This study emphasises the Covid-19 consequences on the semiconductor and automotive industries and the effect of the increase in crude oil prices on the semiconductor industry. To provide an answer to the research question, a cointegration analysis was performed on five estimated models. Therefore, this paper provides an overview of the literature on semiconductor supply chain disruptions and their effect on the automotive industry and analyses the potential impact of the increase in crude oil prices on the volume of the semiconductor supply chain.

**Keywords:** semiconductor industry, automotive industry, supply chain disruptions, China, the U.S., Covid-19, pandemic, economic effects, crude oil, motor vehicles, semiconductors, impacts

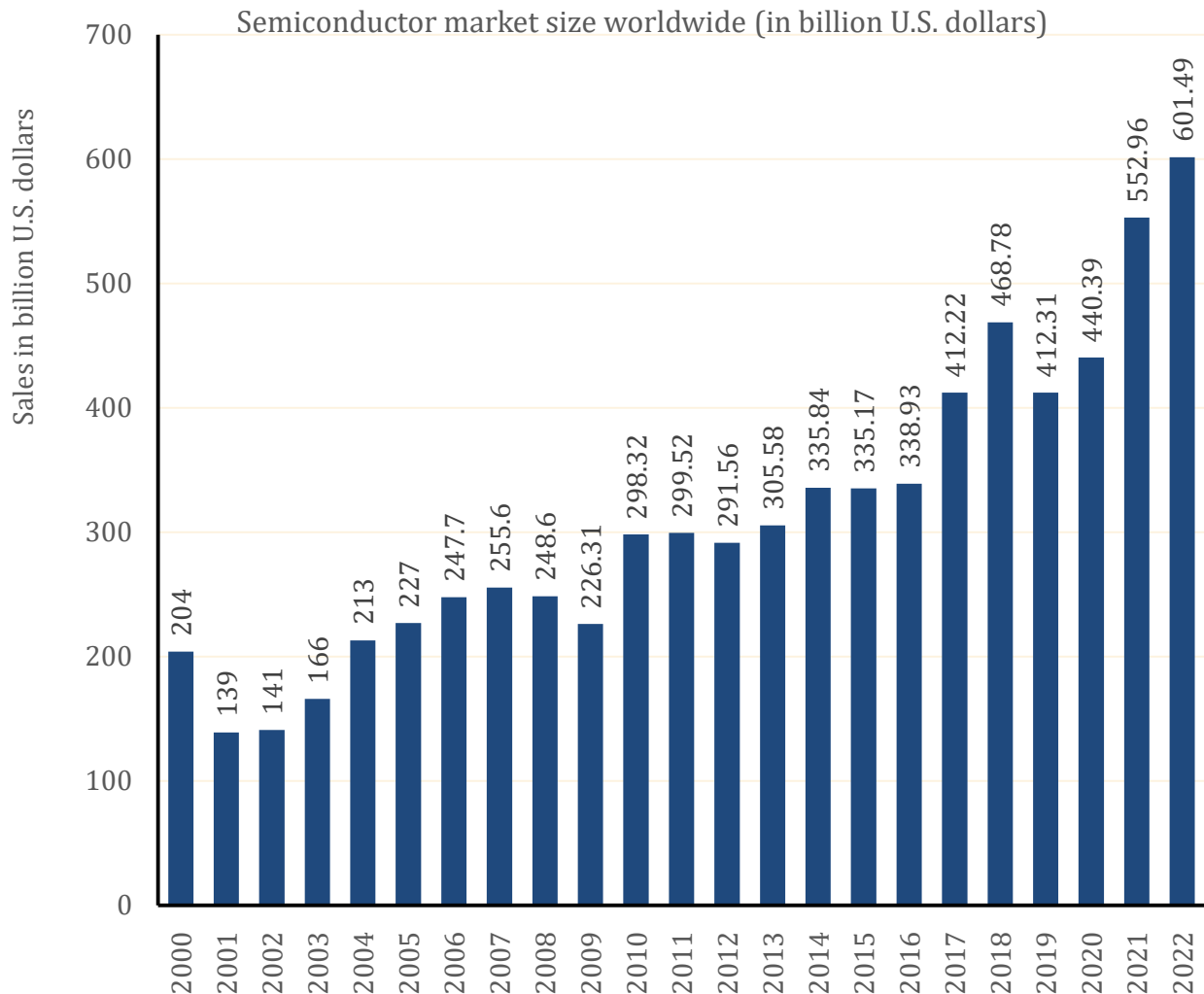
### **1. INTRODUCTION**

#### **1.1 BACKGROUND INFORMATION**

Over the years, the application of semiconductors, also known as chips, which are tiny electronic components that function as hybrids of an insulator and a conductor, became widely regarded as an essential building block of modern industrial and national security technologies. Semiconductors are primarily implemented to produce or finalise the production of many different technological devices and goods, including automobiles. In order to be efficient and functioning, the semiconductor supply chain, as defined and described by TSMC, Taiwan Semiconductor Manufacturing Company Limited (2022), comprises and involves plenty of procedures and various steps in the procurement process, production, and fulfilment process. Additionally, TSMC (2022) points out that to reduce the environmental impact semiconductor production has and promote sustainability, the semiconductor supply chain management should improve the manufacturing processes and the quality of the products. Especially since digital advancements and the use of semiconductors in the contributions to quality technological progress and its implementation are essential, its environmental impact is vital for sustainability and further economic improvements and developments in the semiconductor and automotive industries, respectively. Digital technologies are changing traditional business models, and new opportunities and new business environments are provided with Industry 4.0 or the Fourth Industrial Revolution-the Fourth Industrial Revolution refers to the convergence of biological, physical, and digital innovations, intending to strive toward achieving and converting to cyber-physical systems (Schwab, 2021)- giving intelligent automation and interconnectivity much importance and leaving opportunities for the industrial growth of the semiconductor industry since most electronic devices require the installation of semiconductors in the production process.

The semiconductor market size has grown substantially from a global perspective in the past twenty years. According to Statista (2022), Figure 1 shows the worldwide rising trend of semiconductor market size. Over the last twenty-two years, the market size has noticeably increased by 294.85 per cent in 2022, with sales of 601.49 billion U.S. dollars compared to 2000 with 204 billion U.S. dollars sales. With this point being made, any disruptions in the supply chain of semiconductors will immensely affect other industries since, in most industries, during production, semiconductors are one of the main components. During the production process in the automotive industry, semiconductors are critical, and the disruptions in the supply chain will halt the production of automobiles.

Figure 1: Semiconductor market size worldwide from 2000 to 2022 (in billion U.S. dollars)

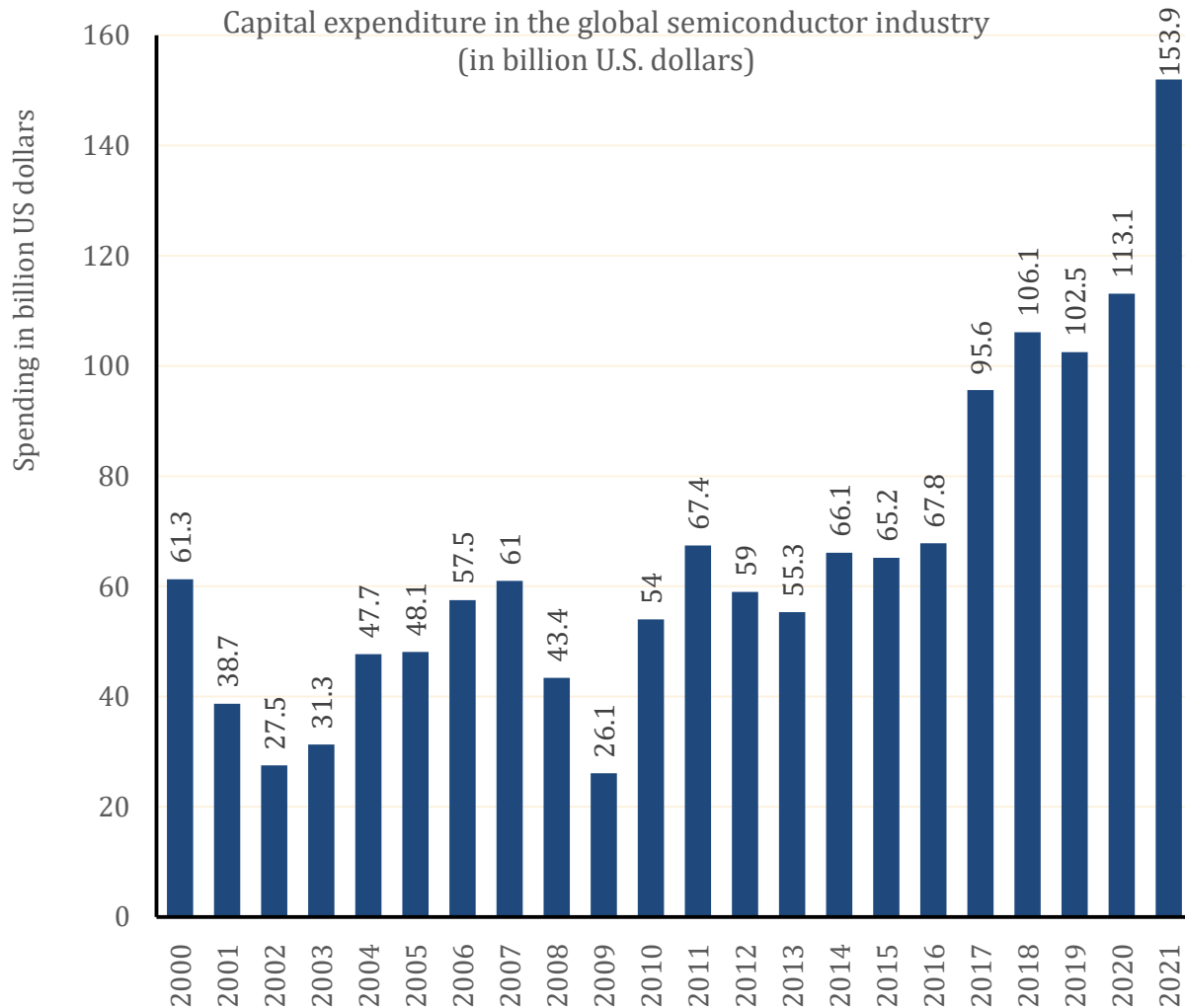


Source: Statista, 2022

According to the data released by Statista (2022), the global capital expenditure in the semiconductor industry was the lowest at the end of the Great Recession in 2009, amounting to 26.1 billion U.S. dollars in spending, which was about 42.4 per cent of the industry's capital expenditure in 2000, which amounted to 61.3 billion U.S. dollars. As the economy recovered, capital expenditure increased, and the most significant increases, compared to the capital expenditures in the previous year, happened in 2010, 2017, and 2021.

In 2021, the global capital expenditure in the semiconductor industry, which was the highest increase percentage-wise -a jump of 36 per cent- since 2000, amounted to 153.9 billion U.S. dollars. The semiconductor market size is consistent with the capital expenditure in the semiconductor industry.

Figure 2: Capital expenditure in the global semiconductor industry from 2000 to 2021 (in billion U.S. dollars)



Source: Statista, 2022

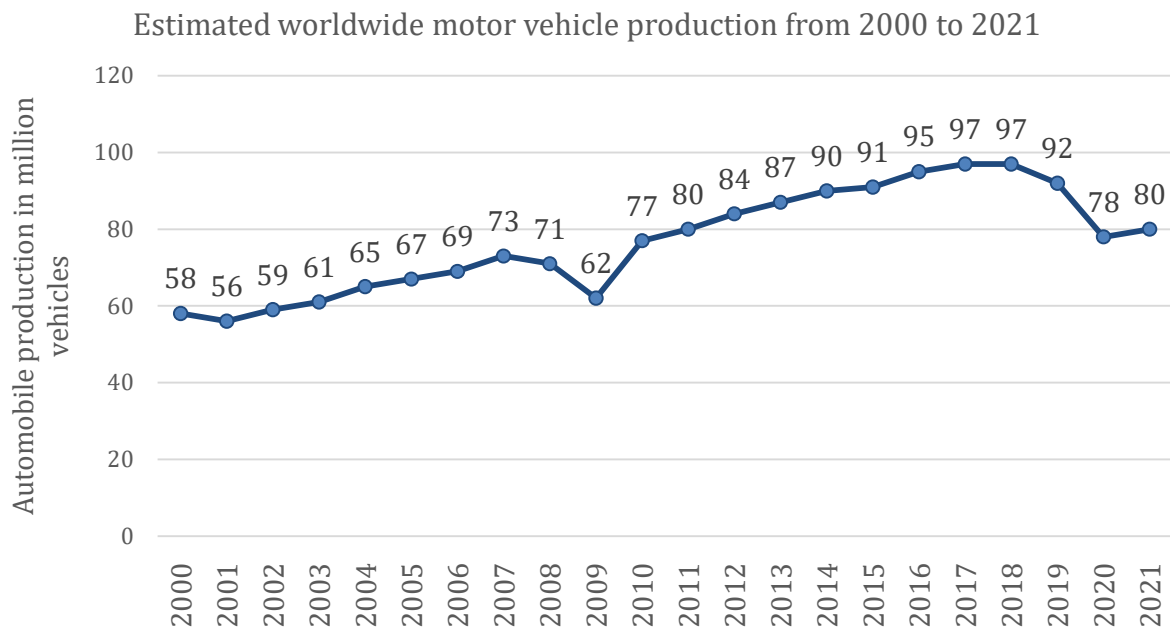
## 1.2 AUTOMOTIVE AND SEMICONDUCTOR INDUSTRY IN CHINA AND U.S.

The consistent growth in semiconductor capabilities and performance has boosted economic output and productivity in most countries like the United States of America and China, including 5G, Artificial Intelligence (A.I.), and autonomous electronic automobiles. In particular, the United States has been a leading state in semiconductor research and development, chip design, and semiconductor manufacturing. On the other hand, the semiconductor supply chain comprises the US and several foreign companies from Europe and the Middle East that control fabrication facilities (Min & Jianwen, 2020).

While the Semiconductor industry in the US declined due to its reliance on global supply chains from East Asia and the Covid-19 pandemic, China has witnessed an emerging increase in power and stability in the market size for semiconductors due to state-led efforts, thus posing a high level of competition to the United States.

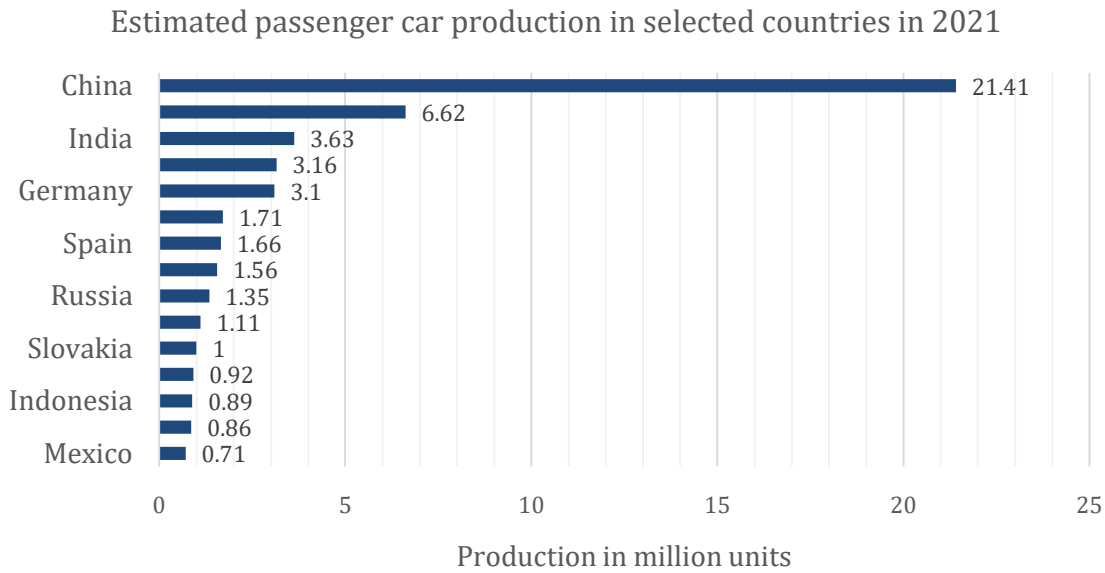
Statista (2022) released data regarding estimated worldwide motor vehicle production from 2000 to 2021; as we can notice in Figure 3, in 2021, the vehicle production amounted to 80 million worldwide produced vehicles, which increased about three per cent when compared to 2020. Predating the Covid-19 pandemic, in 2018, the estimated worldwide motor vehicle production was about 97 million produced vehicles worldwide. Two significant drops in the production of motor vehicles in the last twenty years occurred; the first drop was during the Great Recession from 2008 to 2009, and the second one was during the Covid-19 Pandemic from 2019 to 2020. Additionally, Statista Research Department (2022) estimated countries' ranking regarding the estimated production of passenger cars, four-wheeled motor vehicles that are meant for transporting passengers and have no more than nine seats when the drivers' seat is included, and the leading representative country in 2021, as we can see in Figure 4, was China, producing about 21.4 million units and amounting to almost a third of the passenger vehicle production in the world, and emerging as the globally primary growth market zone for automotive industry players. The U.S. was placed eighth, with the production amounting to 1.56 million units in 2021.

Figure 3: Estimated worldwide motor vehicle production from 2000 to 2021 (in million vehicles)



Source: Statista, 2022

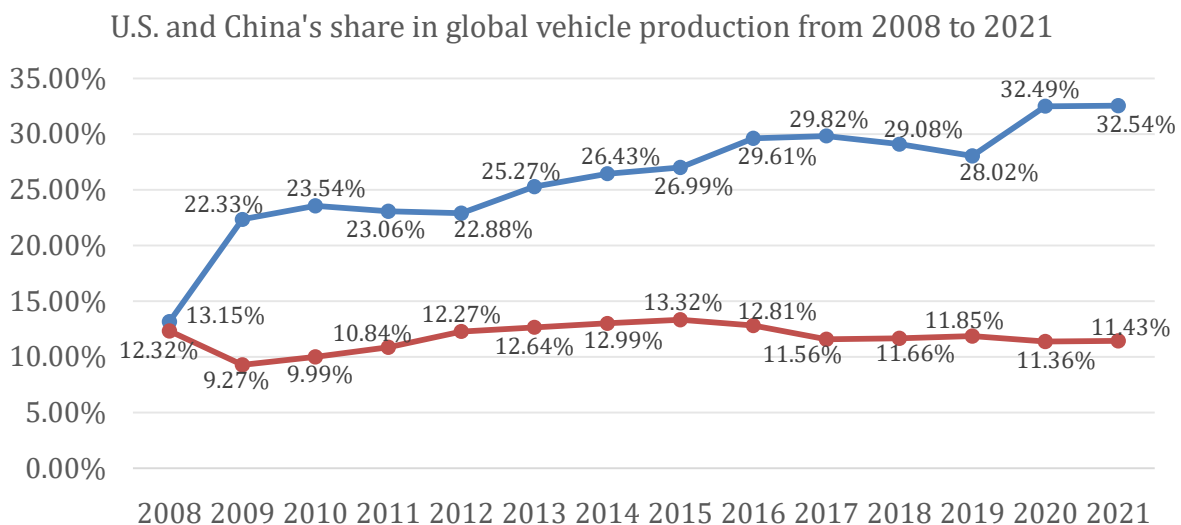
Figure 4: Estimated passenger car production in selected countries in 2021(in million units)



Source: Statista, 2022

Figure 5 is a representation of China's and U.S. share in global vehicle production from 2008 to 2021. Vehicle production of the U.S. expressed in a fraction of the vehicle production worldwide, in 2008 amounted to 12.32 per cent, while China's was 13.15%. They were in a close proximity. Recently, in 2021, U.S. production was equal to 11.43 per cent of global share, China's share in global vehicle production in 2021 amounted to 32.54 per cent. China has significantly increased its share -China's share in global vehicle production almost tripled- in global vehicle production since 2008, while the U.S. share in 2021 was lower -U.S. share in global vehicle production was almost one per cent lower than it was in 2008. The production of motor vehicles in the U.S., when calculated as the fraction of worldwide production, has halved in the span of 22 years. There are plenty of possible explanations, and one of those would be semiconductor supply chain disruptions and chip shortages.

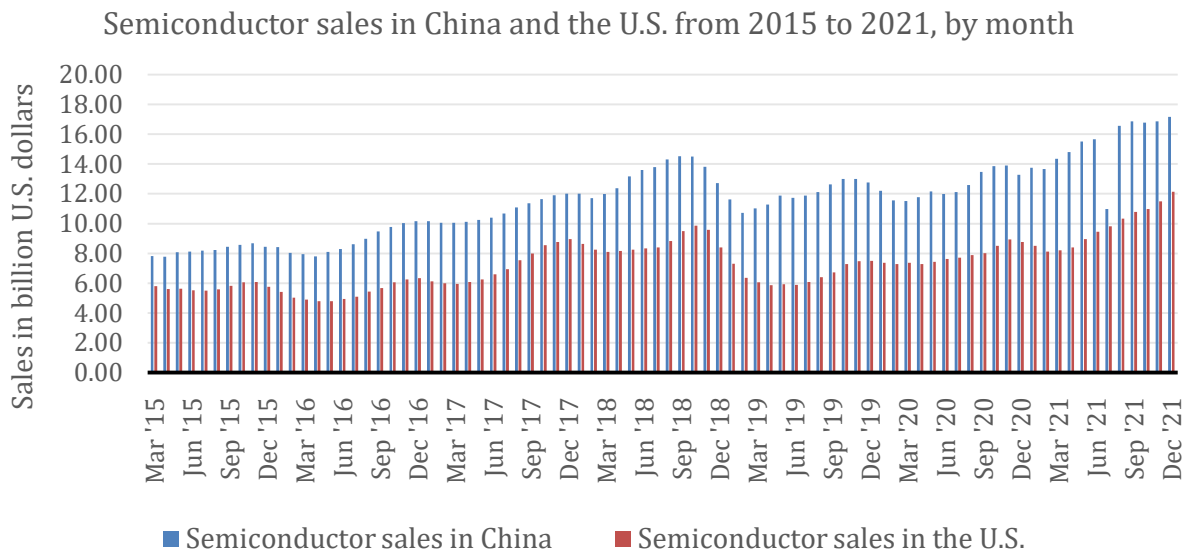
Figure 5: U.S. and China's share in global vehicle production from 2008 to 2021.



Source: Statista, 2022

When we observe semiconductor sales in China and the U.S. from 2015 to 2021, as the Figure 6 dictates, China's sales are larger than the U.S. semiconductor sales. In December 2021, Statista(2022)reported that China's semiconductor sales reached 17.16 billion U.S. dollars, while the U.S. sales reached 12.14 billion U.S. dollars. According to Statista (2022), worldwide, total semiconductor sales were estimated to be 50.85 billion U.S. dollars in 2021.

Figure 6: Semiconductor sales in China and the U.S. from 2015 to 2021, by month (in billion U.S. dollars)



*Source: Statista, 2022*

Since attention is paid to maximising profits and increasing market share additionally, more than before, the semiconductor global value chain is facing some challenges of vulnerability(Boquet, 2021).

### **1.3 PROBLEM STATEMENT**

The semiconductor industry is very competitive since, as technology changes consistently through innovation, competitors tend to bring the latest designs into the global market. The global supply chain for semiconductors comprises the U.S., Japan, South Korea, Taiwan, and China, which has experienced immense growth in the industry, accounting for a more significant percentage of global demand for semiconductors. However, some changes in the supply chain of semiconductors were applied due to the COVID-19 pandemic due to increased demand for specific electronic devices, resulting in supply shortages of semiconductors in the automotive industry. Therefore, this study tries to identify the factors affecting the semiconductor industry, its supply chain, and the emergence of semiconductors in China and the United States of America while defining difficulties in the automotive industry that arose due to semiconductor supply chain shortage alongside providing the recommendation for automotive companies to remaining competitive in the industry.

### **1.4 RESEARCH QUESTION**

Given the information provided so far in the introduction, the primary motivation behind this study is to examine the following questions of *how the semiconductor industry has been affected in recent years, what caused the supply chain disruptions and how that reflected on other industries, specifically the automotive industry, correspondingly answering the question how changes in semiconductor industry affect changes in the automotive industry.*

Additionally, particularly considering recent events of the COVID-19 pandemic and its influence on the semiconductor and automotive industry are considered in this research, as well as the increase in crude oil prices, which is directly connected to the manufacturing process since the equipment used functions on it. Moreover, emphasising how the semiconductor supply chain disruptions affect the automotive industry in China and the U.S., how to cope with these disruptions, the associated challenges, and the direction of the semiconductor and automotive industries are discussed.

### **1.5 RESEARCH AIMS AND OBJECTIVES**

This research aims to ascertain and discuss the causes of the semiconductor supply shortage and how the shortage affected the semiconductor industry. In addition, it analyses the effect of chip manufacturing shortage on the production and global distribution of automobiles. The research provides an outline for approaching the semiconductor industry's future perspective of resolving supply chain disruptions and the prospect of implementing electronic automobiles considering the semiconductor industrial possibilities and opportunities in the long term.

Current disruptions in the supply chain of the semiconductor industry that present an impactful element for lately occurred challenges in the automotive industry have been caused by several factors, including increased demand in both industries, recent events of the COVID-19 pandemic, the increased price of crude oil, and the deficiency of labour capacities, which affects the undertaking for implementation of electric vehicles (E.V.s) and impacts the planned course for reduction of greenhouse gas emissions.

### **1.6 THESIS STATEMENT**

The Covid-19 pandemic, measures implemented to recover the economy and industries' capacities, and changes in crude oil prices caused the semiconductor supply chain disruptions, decreasing worldwide motor vehicle sales and increasing China's share in global vehicle production in the long run.

## **2. LITERATURE REVIEW**

### **2.1 SEMICONDUCTOR INDUSTRY AND ITS KEY SUCCESS FACTORS**

Accomplishing success in the semiconductor industry is influenced by some key factors. The most critical factors for the development of companies in the semiconductor industry are (1) R&D expenditures, (2) governmental aids and policies, (3) the innovation cycle, (4) collaboration and cooperation, (5) the level of dependency on foreign countries, (6) education and workforce, (7) restrictions to foreign competitors, (8) demand from space industry and defence. (Boquet, 2021)

*Research and development (R&D) expenditures*– since the semiconductor industry is constantly innovating, and technological innovations and industries require a creative approach to developing and designing better and more efficient products, one of the critical success factors in the semiconductor industry is R&D and expenditures on R&D. Chinese companies have the lowest percentage of investments in R&D in terms of percentage of GDP; however, they have experienced the highest growth rate of investments in information and communications technologies (ICTs). Both U.S. and Chinese companies are investing in information and communications technology. Manufacturing equipment holds tremendous importance regarding the produced quality and quantities in research and development as a success factor. One can recognise how well have companies implemented R&D strategies based on the number of innovations regarding development or discoveries; based on the number of granted patents. (Boquet, 2021)

*Governmental aids and policies*– are vital because they allow companies to invest more in innovation due to reduced costs. They are provided in the form of new program implementations, financial incentives, and the provision of new policies and standards for enhancing the national market.

In the U.S., research departments, such as the Defence Advanced Research Projects Agency (DARPA), the United States Department of Energy (DOE), and the National Institute of Standards and Technology (NIST) were provided with subsidies and investments since they promote and pay attention to innovations and technological progress, boosting the country's position in the technological industry, as well as the country's technological security. The U.S. government introduced the "Chips for America Act" as federal support for expanding capacities within semiconductor manufacturing companies. Since the U.S. plans to relocate their companies in China to the U.S., and because relocating costs would be massive, the government's provision of sufficient incentives is essential for companies to endure the relocation difficulties and not be discouraged. However, in China, the government is also providing incentives and aid and subsidies for manufacturing equipment, especially since they emphasized the semiconductor industry's importance and implemented "the medium- and long-term plan for Science and Technology". (Boquet, 2021)

*The innovation cycle*—as in any industry, the semiconductor industry places enormous importance on innovation, but innovation is not a sporadic process or just some phase of a particular process and therefore requires constant investment to provide better qualities and improved technologies to initiate the innovation cycle. Its importance additionally lies on and is also reflected in providing opportunities for future generations to implement innovative approaches in the semiconductor industry. (Boquet, 2021)

*Collaboration and cooperation*—for efficient industrial development, cooperation and collaboration with economic entities are the required factors for success in the semiconductor industry. Both the U.S. and China have substantial levels of collaboration and cooperation. (Boquet, 2021)

*The level of dependency on foreign countries*—for semiconductors to be produced, the involvement of several countries is necessary since the production of semiconductors is like a global ecosystem. The amounts of imports and exports can determine the levels of interdependency; lower exports mean higher economic and manufacturing interdependence, and lower imports mean lower economic and manufacturing interdependence. Additionally, since oil is necessary for the manufacturing equipment required for the production of semiconductors, the dependency on foreign countries that export oil is high, especially for China, because the U.S. is one of the countries that export oil. (Boquet, 2021)

*Education and workforce*—given that education and training play a significant role in the semiconductor industry because the semiconductor industry is based mainly on applying knowledge to make and achieve innovations, the country's level and the possibility of education are fundamental. Since a country's education is correlated to the number of provided universities, in the U.S., there are eight universities of engineering and technology that are ranked within the top twenty-five in the world, while in China, there are only two of the same rank. (Boquet, 2021)

*Restrictions to foreign competitors*—to protect the domestic market from the tactical aggression of competition from foreign companies, it is necessary to set restrictions and controls at the federal or regional level. Due to geopolitical tensions, both China and the United States have imposed stricter and increased regulations and controls to protect their markets. In the U.S., the Entity List was established, and in China, the Unreliable Entity List was also released alongside Cybersecurity Law. (Boquet, 2021)

*Demand from the space industry and defence*—investments in the military sector for better defence and the space industry increase the research and development resources of the semiconductor industry since semiconductors are crucial components for the utilised equipment - it is, however, inconclusive that the higher the spending for the military sector, the higher the growth within the semiconductor industry. (Boquet, 2021)



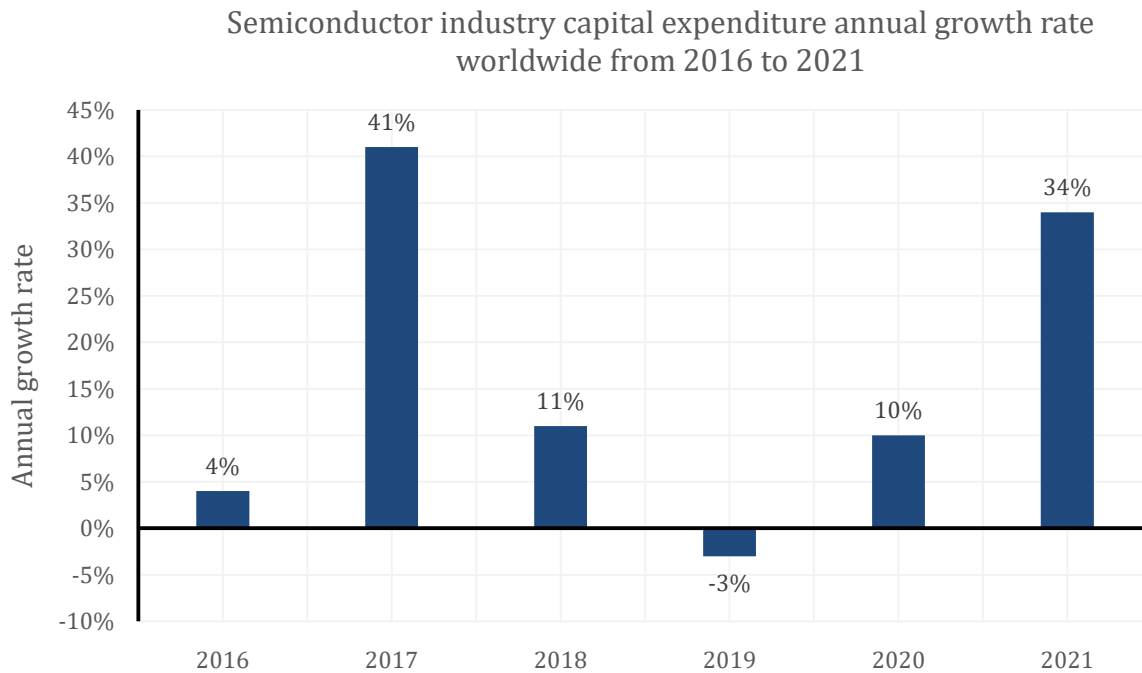
## **2.2 IMPACT OF COVID-19 ON THE SEMICONDUCTOR INDUSTRY**

Due to the Covid-19 pandemic between 2020 and 2021, the demand for semiconductors significantly increased, resulting in global chip shortages (Wu, et al., 2021). During the pandemic, chipmakers saw the surging demand for semiconductors in various sectors such as health, virtual learning, and other automotive sectors. As a result, the semiconductor industry witnessed a projection in the expenditure in the following year of 2021 to 152 billion U.S. dollars from around 113 billion U.S. dollars in 2020, as illustrated in Figure 1. Prior to 2021, the industry had never spent above 150 billion U.S. dollars on annual capital expenditure, and the need for the global semiconductor industry to meet this projected market growth in the years ahead has prompted industries to not only increase the fabrication capacity alone but also other sectors like the transportation sector. Generally, as the demand for electronics and automobiles continues to grow, the demand for chips will also continue to rise, leading to an increase in annual capital expenditure in the semiconductor industry. While the 2020 markets forecasts fluctuated throughout the year due to demand uncertainty resulting from the Covid-19 pandemic, the sales volume for semiconductors increased in 2020, forecasting further growth in the coming years. As the semiconductor industry in China is one of the fastest-growing industries globally, a sales increase of 16% accommodated the leading Chinese semiconductor companies, despite the Covid-19 pandemic and the global sales increase of 5.8% (Boquet, 2021).

Covid-19 influenced customer behaviour, corporate operations, and companies' revenues in the semiconductor industry. Unpredictability and uncertainty due to the pandemic have caused difficulties in semiconductor companies' strategic decision-making. Bauer et al. (2020), senior partners of McKinsey & Company, a trusty management consulting firm founded in 1926, conducted a study to develop recovery strategies for the semiconductor industry to emerge stronger after the Covid-19 crisis. As the market recovers, the situation regarding revenue growth, which is experiencing negative revenue growth, is expected to improve—trends of videoconferencing open new possibilities for innovations in the semiconductor industry. The semiconductor industry can become stronger after the Covid-19 crisis by (1) reducing capital expenditure, (2) focusing on R&D, and (3) strategically approaching acquisitions and mergers. Uncertainty in the market and creating possible future scenarios that show different outcomes are essential for formulating strategies for the upcoming years—the uncertainty should be embraced as part of the operation management model because adapting quickly to the situations will be of greater importance following long-term plans. (Bauer, et al., 2020)

When we observe the semiconductor industry capital expenditure annual growth rate worldwide from 2016 to 2021, as the Figure 7 shows, we can notice that there is a significant noticeable change in the year when the Covid-19 pandemic started; the industry had a positive growth rate until 2019 when a negative growth rate of three per cent was noted. We can conclude that the pandemic impacted the amount of capital expenditure in the industry and that the following increase in capital expenditure growth rate is due to increased demand and supply shortages to recover the industry's supply as to demand requirements.

Figure 7:Semiconductor industry capital expenditure annual growth rate worldwide from 2016 to 2021



Source: Statista, 2022

### **2.3 IMPACT OF THE SEMICONDUCTOR SUPPLY CHAIN AND COVID-19 ON THE AUTOMOTIVE INDUSTRY**

Automobile supply chains have developed, are dispersed geographically and are more complex when compared to three decades ago; however, unpredictable disruptions caused difficulties with automobile supply chain management. The World Trade Statistics (2021) project that a significant 527 billion U.S. dollars increase in total sales of semiconductors is likely to be realised. These projections only show that the future of the automotive industry is majorly reliant on the production of chips in the semiconductor industry, which has been witnessed to be inadequate. Therefore, research and development sectors in the semiconductor industry are left with a great responsibility for production to meet the surging demand for the chips.

Companies have problems with delivery plans in the automotive industry because of the poor planning environment. Johansson and Westmark(2021) designed new processes for handling orders and freeze periods to decrease the volatility within delivery plans. Four pilot customers were tested with methods for implementing the new process. Minor changes and easily interpreted processes created a more straightforward implementation of the new processes. Automation saves time; however, it is vital to check all deviations manually. It is also essential to arrange meetings with customers, reach agreements, and implement contracts for freeze periods. The pilot experiment resulted in customers reducing the number of changes and agreeing on new freeze periods. The limitations of their experiment are that the sample size was too small, and the duration of the experiment was limited. When volatility is high, making decisions and planning the production is complicated and challenging, increasing expenses, creating inventory imbalances, and questioning quality. However, the limitation of their study is that the duration of the experiment was too short. The opportunity to meet with the companies' staff who participated in the experiment was disabled due to the pandemic. The occurrence of the semiconductor crisis disturbed data.

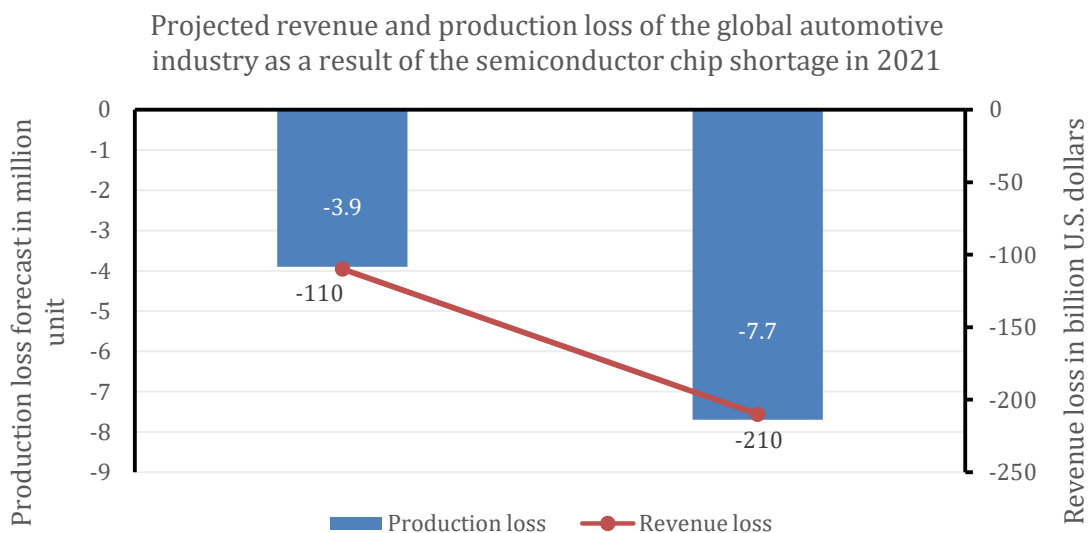
If agreements on freeze times are achieved and customers are included in that decision, volatility assurance will be lessened, and processes will be improved.(Johansson & Westmark, 2021)

Jiang, Shu, and Miao. (2021), Professors at Sichuan Normal University, University of Electronic Science and Technology of China, and The Hong Kong Polytechnic University, who have expertise in mathematics, management, and logistics, have conducted a study to propose an optimal plan of production of vehicles by substituting each chip and sourcing for maximising the total revenue. A deterministic mixed-integer programming model is proposed, and a min-max robust model is constructed based on the deterministic model. A two-stage stochastic programming model is defined. A set of numerical experiments is conducted to demonstrate the advantage of the min-max robust model compared to the deterministic model. The robust model proposed in the study prohibits disturbances from uncertain parameters. The results of calculations for the robust model, when compared with the other two proposed models, are not negligible.(Jiang, et al., 2021)

Nayak et al. (2021), Professors at different universities in India that have expertise in electronics and information technology, have conducted a research to study Covid-19 impact on six different industries, approaching the automotive industry in detail – the study seeks to provide a base for overcoming the crisis in the automotive industry regarding purchases. Purchase decisions were affected by household income, the pandemic's effect on local regions, and travel difficulties. Lockdowns decreased immigrant labour, on which the automotive industry relies – the welfare of active workers and communication with them is essential. The disruptions of China's exports of parts for automobiles affected the U.S. the most, and the production of automobiles declined because of continuous lockdowns that influenced consumer and producer demand. Evidence of vehicle sales before and after Covid-19 and empirical analysis of scientific studies and models supports the study's claim. (Nayak, et al., 2021)

When we observe the projected revenue and production loss of the global automotive industry as a result of the semiconductor chip shortage in 2021, as the Figure 8 illustrates, we can see that the projected revenue loss amounted to 250 billion U.S. dollars and that the production loss was forecasted to be 7.7 million units of production. The industry suffered huge losses due to the Covid-19 pandemic, and they have increased since the beginning of the pandemic.

Figure 8: Projected revenue and production loss of the global automotive industry as a result of the semiconductor chip shortage in 2021(in billion U.S. dollars and million units)



Source: Statista, 2022

Digital transformation can, however, provide the advantage of the faster economic recovery of the automotive industry after or during the Covid-19 crisis. To remain competitive in the industry, automotive companies should invest in and adapt to digital transformation to be more productive, competitive and have more profit - companies should invest in research, innovation, and development activities.(Llopis-Albert, et al., 2021)

#### **2.4 CHALLENGES AND OPPORTUNITIES**

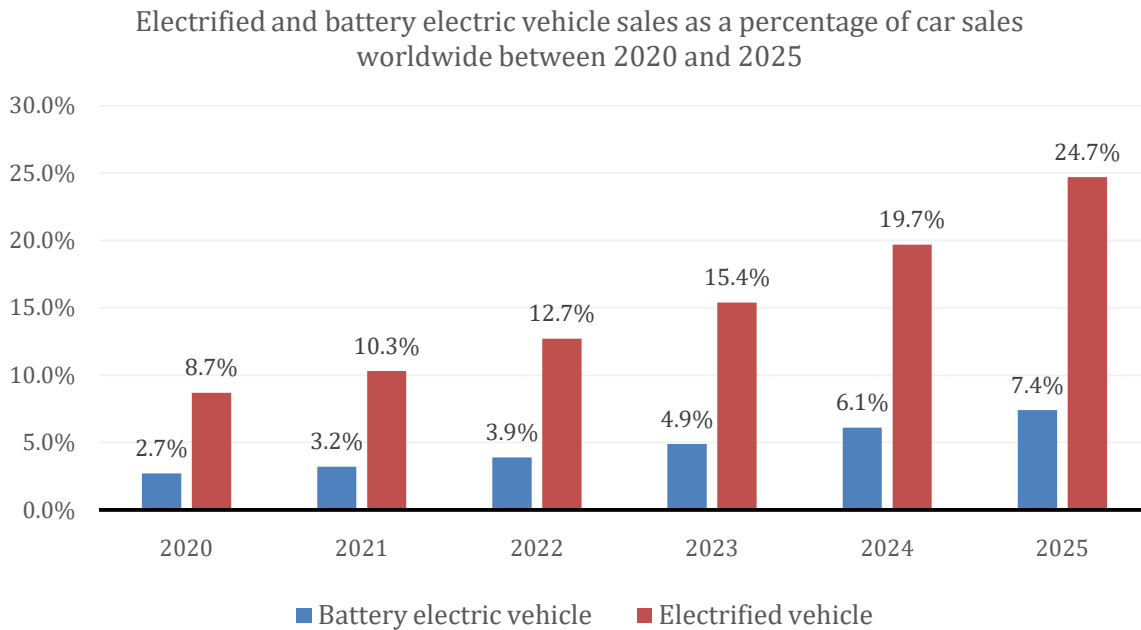
Unpredicted events of the Covid-19 pandemic affected the semiconductor supply chain, which was impacted by challenges and provided opportunities. Mainova and Bitri(2021), Professors at universities in Bulgaria and Albania who have expertise in technologies and management, have conducted a study to identify solutions applied to the automotive industry to issues caused by the pandemic in the semiconductor supply chain. During the Covid-19 pandemic, since the increased demand for electronic-based products increased the demand for semiconductors because of their necessity in the manufacturing, the lockdowns and implemented measures decreased and slowed down the production of semiconductors, which had a chain effect and created challenges in industries due to the supply shortages. For the automotive industry, an additional increase in demand besides semiconductor supply shortages caused multiple difficulties, especially in the green car market (e.g., Tesla Inc.), since green cars or electric vehicles rely solely on semiconductors, among other elements in the manufacturing phase. China and the U.S. have implemented initiatives to overcome or make the most out of the situation. (Marinova & Bitri, 2021)

Three responses to Covid-19 and chip shortage were noted, including how countries reorganised their approach to establishing local and independent production of semiconductors, with an emphasis on China. The instability in the semiconductor supply chain was due to inadequate decision making; allocation of resources was not conducted based on requirements, which opens opportunities for new companies to take the lead. This instability, however, including the lack of local opportunities in China, has increased the technological and trade war between the U.S. and China, which without investments in the semiconductor industry, provides a challenge for China's automotive industry, especially since China's government placed a regulation that the country will fully converge to electric vehicles by 2030(Hill, et al., 2020).This unstable situation led China to encourage investment from foreign companies to bring their research centres, technologies, and talents to China. Consequently, the U.S. is concerned about China's potential leadership in the semiconductor industry due to their foreign investment attraction policies.(Marinova & Bitri, 2021)

However, the situation regarding the semiconductor supply shortage was worsened due to the cold wave in Texas and a fire at the factory in Japan—Toyota, Volkswagen, and G.M. cut their production. In response to the supply shortage of automotive semiconductors, the U.S. Biden administration promised support for semiconductor manufacturers. It is necessary to build a cooperative network between the automotive and semiconductor industries to expand development and production capabilities in the long term.(Chun, et al., 2021)

When we observe electrified and battery vehicle sales as a percentage of car sales worldwide between 2020 and 2025, as Figure 9 displays, the estimation indicates that the demand is increasing both for electrified vehicles and battery electrified vehicles. This forecasted increase means the industry will open new opportunities for the semiconductor and automotive industries separately and collaboratively - the demand for semiconductors will be boosted since the production capacities will have to be met. More production capacities in terms of quantities and innovations are required to prevent the prolongation of semiconductor supply shortages.

Figure 9: Electrified and battery electric vehicle sales as a percentage of car sales worldwide between 2020 and 2025



*Source: Statista, 2022*

## 2.5 LIMITATIONS

Potentially, an abundance of subjective and distorted information is currently present online and in other formats due to the existing geopolitical tensions between China and the U.S. (Boquet, 2021). Additionally, for the literature review, no studies were found to cover the impact and correlation of crude oil price increases on production capacities within the semiconductor industry. It is essential to note that data used for further analyses and obtained from Statista, one of the leading companies on the internet specialising in consumer and market data, was mainly available for different periods within the range 2000 to 2025.

## 3. METHODOLOGY AND DATA

Data explanation was presented as the first phase of the methodology section. For data explanation, descriptive statistical analysis was conducted in Stata. In the following phase, stationarity was checked by conducting a unit-root test using the augmented Dickey-Fuller (ADF) test. Conversion of the resulted variables into the first-differenced form was performed. Afterwards, a second difference was also taken. After that, to determine the presence of cointegration between the variables, Engle and Granger's (1987) two-step approach was implemented to test the long-run relation and connection between worldwide semiconductor unit shipments and worldwide motor vehicle sales, China's share in global vehicle production, average annual West Texas Intermediate crude oil spot prices, average annual Europe Brent crude oil spot prices, and average annual OPEC crude oil price respectively. *Cointegration* is an analysis that tests for and explains the long-run relationship of correlation between time series variables (Wooldridge, 2013).

### 3.1 DATA EXPLANATION

For the purpose of this study, data on semiconductor unit shipments worldwide were used for the period from 2000 to 2025 (with estimates from 2022 to 2025). Data on worldwide motor vehicle sales was used from 2005 to 2021. Data regarding China's share in global vehicle production was used from 2008 to 2021.

Data regarding average annual West Texas Intermediate crude oil spot prices and average annual Europe Brent crude oil spot prices were used for 2000 to 2021, while data on average annual OPEC crude oil price was used for 2000 to 2022. The differences in the period were due to the unavailability of data.

The source of the statistical data is the publications of Statista, one of the leading companies on the internet that specializes in consumer and market data.

Descriptive statistics of each variable included in the econometric analysis -stationarity and cointegration- are presented in the following table:

Table 1: Descriptive statistics of variables

NAME	DESCRIPTION	MEASURE	OBS	MEAN	STD. DEV.	MIN	MAX
semiconductorshipment	semiconductor unit shipments worldwide	in billions	26	780.30	291.69	332.90	1305.70
wwmotorsales	worldwide motor vehicle sales	in million units	17	80.93	10.49	65.57	95.66
chinashare	China's share in global vehicle production	percentage of global vehicle output	14	0.26	0.05	0.13	0.33
westtexas	average annual West Texas Intermediate crude oil spot prices	in U.S. dollars per barrel	22	61.06	24.33	25.60	99.67
europébrent	average annual Europe Brent crude oil spot prices	in U.S. dollars per barrel	22	63.90	28.10	24.00	111.63
opecprice	average annual OPEC crude oil price	in U.S. dollars per barrel	23	63.33	28.03	23.12	109.45

Source: Statista, 2022

### 3.2 METHOD

The method of this study includes:

- testing for cointegration between semiconductor unit shipments worldwide and worldwide motor vehicle sales
- testing for cointegration between semiconductor unit shipments worldwide and China's share in global vehicle production
- testing for cointegration between semiconductor unit shipments worldwide and average annual West Texas Intermediate crude oil spot prices
- testing for cointegration between semiconductor unit shipments worldwide and average annual Europe Brent crude oil spot prices
- testing for cointegration between semiconductor unit shipments worldwide and average annual OPEC crude oil price

Five models are tested on cointegration. First, the analysis begins with checking for stationarity and conducting the augmented Dickey-Fuller (ADF) test to check for unit root. In cases of non-stationarity, the first differences of variables are again tested using the ADF test. Additionally, the second difference of variables is taken and tested using the ADF test for variables that still did not achieve stationarity after the first differences. Engle and Granger's two-step method is applied for the cointegration analyses.

#### 4. RESULTS

The following sections present the process in detail and explain the results of each analysis since the results are better explained while the process is discussed simultaneously.

##### 4.1 STATIONARITY AND AUGMENTED DICKEY-FULLER (ADF) TEST

To start with the process, firstly, all the variables were put in a time-line graph as shown in Appendix (see Appendix A), and it is seen that variables Semiconductor shipments, worldwide motor vehicle sales, and China's global share show a clear trend (upward) pointing towards non-stationarity of these variables. The oil prices, although rising, has strong volatility, and it is not clear if the time series is stationary. For this to formally be confirmed, a unit-root test was initially conducted using the augmented Dickey-Fuller (ADF) test. In the first half of the table, one can look at the variables in their basic form, and the result shows that most of these variables are non-stationary. Conversion into the first-differenced form was performed, and the ADF test was rerun - the result was that stationarity could still not be achieved for some of the variables, when compared to critical t-values. Thus, one step further was performed, the second differences were taken, and results provided that some of the variables now have achieved stationarity.

Table 2: Augmented Dickey-Fuller Test of Stationarity

<b>Variable</b>	<b>t-Value</b>	<b>P-value</b>
Semiconductor unit shipments worldwide	-3.293	0.0674
Worldwide motor vehicle sales	-1.062	0.935
China's share in global vehicle production	-1.347	0.876
West Texas Intermediate	-1.812	0.699
Europe Brent	-1.776	0.716
Average annual OPEC crude oil price	-1.929	0.64
<b>Variable (First Differenced)</b>	<b>t-Value</b>	<b>P-value</b>
Semiconductor unit shipments worldwide	-4.211	0.00431
Worldwide motor vehicle sales	-1.632	0.78
China's share in global vehicle production	-1.766	0.721
West Texas Intermediate	-2.475	0.341
Europe Brent	-2.628	0.267
Average annual OPEC crude oil price	-2.206	0.487
<b>Variable (Second Differenced)</b>	<b>t-Value</b>	<b>P-value</b>
Semiconductor unit shipments worldwide	-4.170	0.00496
Worldwide motor vehicle sales	-2.789	0.201
China's share in global vehicle production	-1.783	0.713
West Texas Intermediate	-3.154	0.0939
Europe Brent	-3.204	0.0836
Average annual OPEC crude oil price	-3.381	0.0540

#### 4.2 COINTEGRATION ANALYSIS

Once it was established that all the variables in the model are I(1) form, the Engle and Granger (1987) two-step approach proceeded. The Engle and Granger (1987) two-step approach includes fitting the model using least squares and estimating the residuals. The models estimated in the first step are given as follows:

$$\begin{aligned} wwmotorsales_t &= \alpha + \beta_1(\text{semiconductorshipment}_t) + \varepsilon \\ chinashare_t &= \alpha + \beta_1(\text{semiconductorshipment}_t) + \varepsilon \\ westtexas_t &= \alpha + \beta_1(\text{semiconductorshipment}_t) + \varepsilon \\ europebrent_t &= \alpha + \beta_1(\text{semiconductorshipment}_t) + \varepsilon \\ opecprice_t &= \alpha + \beta_1(\text{semiconductorshipment}_t) + \varepsilon \end{aligned}$$

In the second step, the residuals were estimated from the above-estimated model, and an ADF was run on the obtained residuals to evaluate for cointegration. The residuals can also be plotted to see the behaviour. Although some of the residuals do not show a cointegrating relationship, estimation of the error correction model (ECM) was still proceeded by running the regression for the following model. The model's coefficients will provide information on the short-run and long-run relationship between the two variables. The following equations were estimated in the second step:

$$\begin{aligned} D2.wwmotorsales_t &= \alpha + \beta_1(D2.semiconductorshipment_t) + \beta_2(L1.Residuals) + \varepsilon \\ D2.chinashare_t &= \alpha + \beta_1(D2.semiconductorshipment_t) + \beta_2(L1.Residuals) + \varepsilon \\ D2.westtexas_t &= \alpha + \beta_1(D2.semiconductorshipment_t) + \beta_2(L1.Residuals) + \varepsilon \\ europebrent_t &= \alpha + \beta_1(semiconductorshipment_t) + \beta_2(L1.Residuals) + \varepsilon \\ opecprice_t &= \alpha + \beta_1(semiconductorshipment_t) + \beta_2(L1.Residuals) + \varepsilon \end{aligned}$$

The interpretation of the above model will provide the short-run and the long-run relationship of the explanatory variable to the dependent variables. The impact of change in short-run semiconductor shipments can be seen in the given and the long-run relationship measured by the lagged value of the residuals estimated in the first stage. The estimates of the five models are given in the following table:

Table 3: Estimates of the five models

	(1) D.Worldwide motor vehicle	(2) D.China's share in global vehicle	(3) D.West Texas Intermediate	(4) D.Europe Brent	(5) D.Average annual OPEC crude oil price
D. Semiconductor Shipment	0.0109 (0.0144)	0.000201*** (0.0000451)	-0.0397 (0.0492)	-0.0547 (0.0528)	-0.0650 (0.0470)
L.Residuals	-0.145 (0.245)				
L.Residuals		-0.979*** (0.166)			
L.Residuals			-0.386** (0.175)		
L.Residuals				-0.356** (0.167)	
L.Residuals					-0.369**



					(0.160)
Constant	1.125 (1.281)	0.0126*** (0.00378)	3.061 (3.942)	3.496 (4.250)	3.871 (3.964)
Observations	16	13	20	20	21

Standard errors in parentheses  
\* p<0.1, \*\* p<0.05, \*\*\* p<0.01

### 1) Semiconductor unit shipments worldwide from 2000 to 2021 & Worldwide motor vehicle sales from 2005 to 2021

The result shows that semiconductor shipment has no short-run or long-run relationship with the worldwide motor vehicle sales. This can be seen from the significance of the coefficient of semiconductor shipment and the coefficient of lagged Residuals value. No interpretation can be made for this relationship.

### 2) Semiconductor unit shipments worldwide from 2000 to 2021 & China's share in global vehicle production from 2008 to 2021

It is found that semiconductor shipments have both a short-run and long-run relationship with China's share in global sales. The result shows that, in the short run, the increase in the share of China in global motor vehicle sales increases semiconductor shipments worldwide. In the long run, this relationship is converging at a rate given by 0.979, i.e., 97% of the disequilibrium will converge in the following time period.

### 3) Semiconductor unit shipments worldwide from 2000 to 2021 & Annual average WTI and Brent crude oil spot prices from 1990 to 2021 (West Texas Intermediate)

It is found that semiconductor shipments have only a long-run relationship with the Annual average WTI crude oil spot prices. The result shows that, in the short run, the crude oil prices as measured by the WTI rate have no impact on the increase in semiconductor shipments worldwide. In the long run, this relationship is converging at a rate given by 0.386, i.e., 38.6% of the difference will converge in the following time period.

### 4) Semiconductor unit shipments worldwide from 2000 to 2021 & Annual average WTI and Brent crude oil spot prices from 1990 to 2021 (Europe Brent)

It is found that semiconductor shipments have only a long-run relationship with the Annual average crude oil price measured by Europe Brent. The result shows that, in the short run, the crude oil prices as measured by the Europe Brent rate have no impact on the increase in semiconductor shipments worldwide. In the long run, this relationship is converging at a rate given by 0.356, i.e., 35.6% of the difference will converge in the following time period.

### 5) Semiconductor unit shipments worldwide from 2000 to 2021 & Average annual OPEC crude oil price from 1960 to 2022

It is found that semiconductor shipments have only a long-run relationship with the Annual average crude oil spot prices measured by OPEC. The result shows that, in the short run, the crude oil prices as measured by OPEC have no impact on the increase in semiconductor shipments worldwide. In the long run, this relationship is converging at a rate given by 0.369, i.e., 36.9% of the difference will converge in the following time period.

## 5. DISCUSSION

### 5.1 IMPLICATIONS OF THE FINDINGS

The aim of this study is to detail the cause of supply chain disruptions in the semiconductor industry and describe its effect on the automotive industry focusing on China and the U.S.; more succinctly said, to demonstrate that (1) the Covid-19 pandemic, together with measures that countries implemented to encourage economy's recovery and industries' capacities improvements affected the semiconductor supply chain disruptions, and that (2) the supply disruptions in the semiconductor industry are also related to crude oil price changes since the equipment used in the production process is dependent on its use, alongside causing a decrease in worldwide motor vehicle sales while increasing China's share in vehicle production in the long run globally.

Although it was stated that the worldwide semiconductor unit shipments were correlated to worldwide motor vehicle sales, the cointegration analysis showed that the relationship between these variables could not be interpreted; thus, one cannot conclude that the semiconductor supply chain is correlated to the sales of motor vehicles. On the other hand, according to the conducted cointegration analysis, the worldwide semiconductor unit shipments were proven to be correlated to China's share in global vehicle production. The analysis showed that the semiconductor shipments increased with China's share growth in global vehicle production in the short run. However, in the long run, the stability of supply, demand, and prices in the semiconductor shipments decreases; in other words, the disequilibrium with the rate of 97% will converge sometime in the future. Regarding the effect of crude oil prices on the supply chain of semiconductors, West Texas Intermediate, Europe Brent, and OPEC were analysed. In the cointegration analysis between semiconductor supply and its distributions and annual average WTI and Brent crude oil spot prices in West Texas Intermediate, it was found that in the short run, there is no increase in the supply shipments, but in the long run, there would be a difference in terms of the relationship converging by 38.6%. The same conclusions were reached for Europe Brent and OPEC; however, the convergence rate for Brent is 35.6%, and for OPEC, it is 36.9%.

These findings may broaden the horizons of researching and analysing the causes of the semiconductor supply chain disruptions to alleviate the industrial and economic impact. This study may help policymakers perceive the importance of good organisational structure and the prospects for developing and innovating other energy sources for manufacturing equipment. With this study, the significance of the semiconductor industry on the economy and automotive industry can be pursued to focus on creating the educational environment as the basis of possible future development of the semiconductor industry in Bosnia and Herzegovina or other developing countries. Since this study focuses on China and the U.S., the limitation regarding econometric analysis is that there were no data available to collect regarding the U.S. share growth in global vehicle production to test for cointegration with semiconductor supply and unit shipment.

## **5.2 RECOMMENDATIONS**

The semiconductor industry is a promising industry that, with the right approach and knowledge, can become prosperous and boost the country's economy, especially in developing countries. Technology is constantly developing, and semiconductors are the core elements of technological innovations. Therefore, it is recommended to invest in making the semiconductor companies more efficient and productive by, for example, employing more people, building more production factories to increase the production capacities, and encouraging younger people to get a good education so that innovations are in perspective as well. Also, in the automotive industry, it is recommended not to rely only on one or few semiconductor suppliers because if the procurement is stopped for some unknown reason or unexpected circumstance, the reliability of companies within the industry is not decreased, and the production of motor vehicles is not halted.

## 6. CONCLUSION

In conclusion, the primary aim of this study was to define and characterise the cause of semiconductor supply chain disruptions and note how that disruption affected the automotive industry while prioritising the outcomes in the industry due to the Covid-19 pandemic and analysing the occurrence of supply chain disruptions due to the increase of crude oil prices since its use for the generation of energy that is obligatory to the manufacturing part is essential. For the research purpose, semiconductor and automotive industries in China and the U.S. were analysed. Critical success factors in the semiconductors industry were provided and described; additionally, the impact of Covid-19 on the semiconductor industry was analysed, and the impact of the semiconductor supply chain and Covid-19 on the automotive industry was explored. The challenges and opportunities were presented and discussed. Additionally, empirical data implied that oil crude prices were significantly correlated with the semiconductor industry and thus indirectly affected the supply chain in the long run. However, the empirical data in Cointegration analysis between the semiconductor supply chain and China's share in global vehicle production could not be interpreted. In this paper, recommendations for the approach to entering the industry for developing countries, and the increase of efficiency in countries that already have entered the semiconductor industry were provided. Additionally, a recommendation for the automotive industry to prevent the halt of production in the future not to rely on one or few semiconductor suppliers was provided. Finally, this research is supposed to expand the point of view regarding the causes of disruptions in the semiconductor supply chain and help understand the importance of creating a prosperous educational environment which is the primary condition for any developing country to enter the semiconductor industry and for countries that already are in that industry to continue to create innovations or to meet the industry's standards.

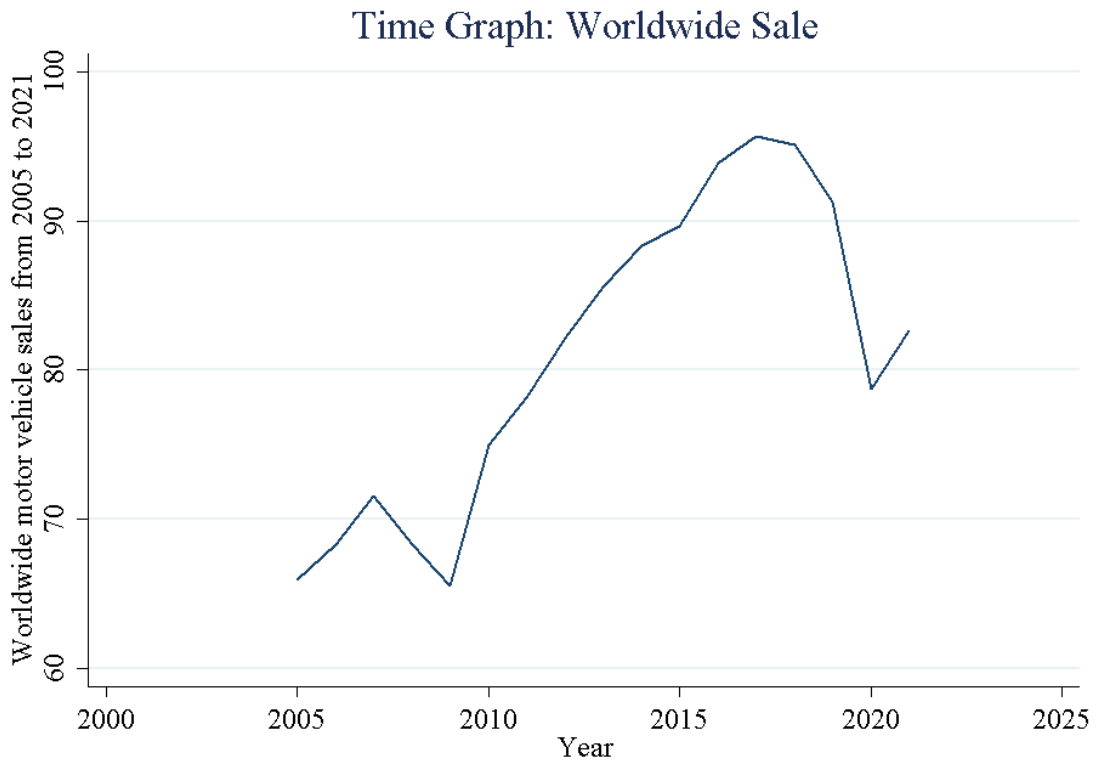
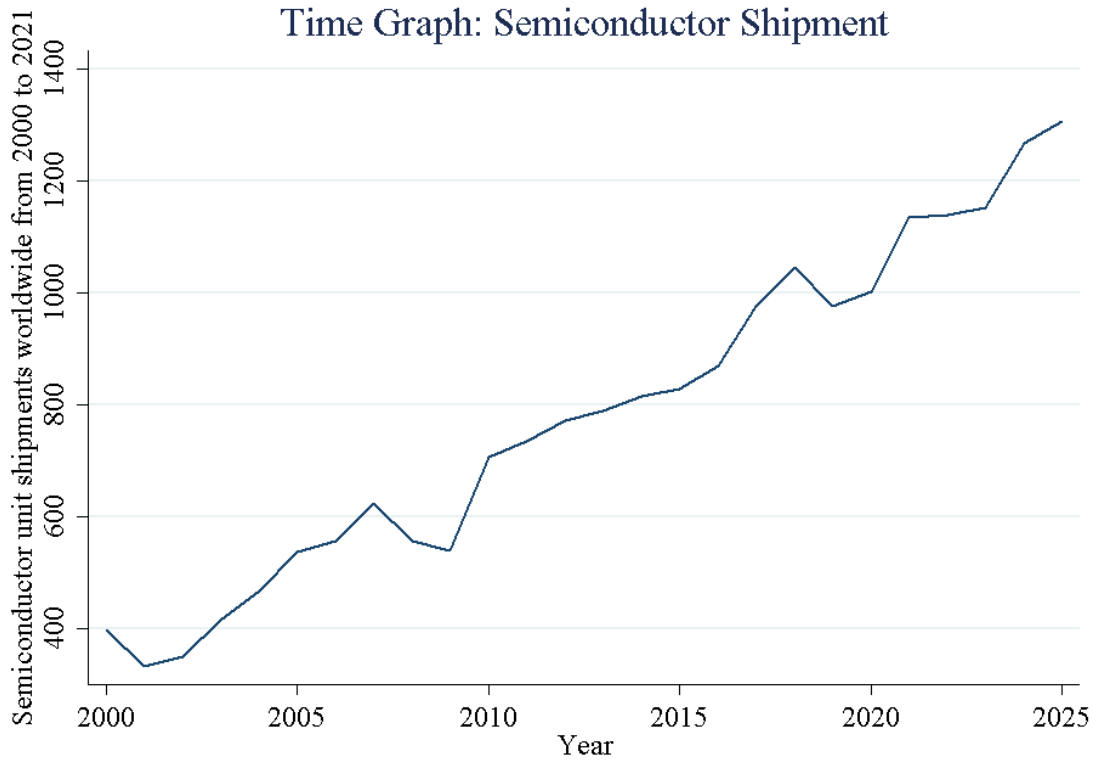
## 8. BIBLIOGRAPHY

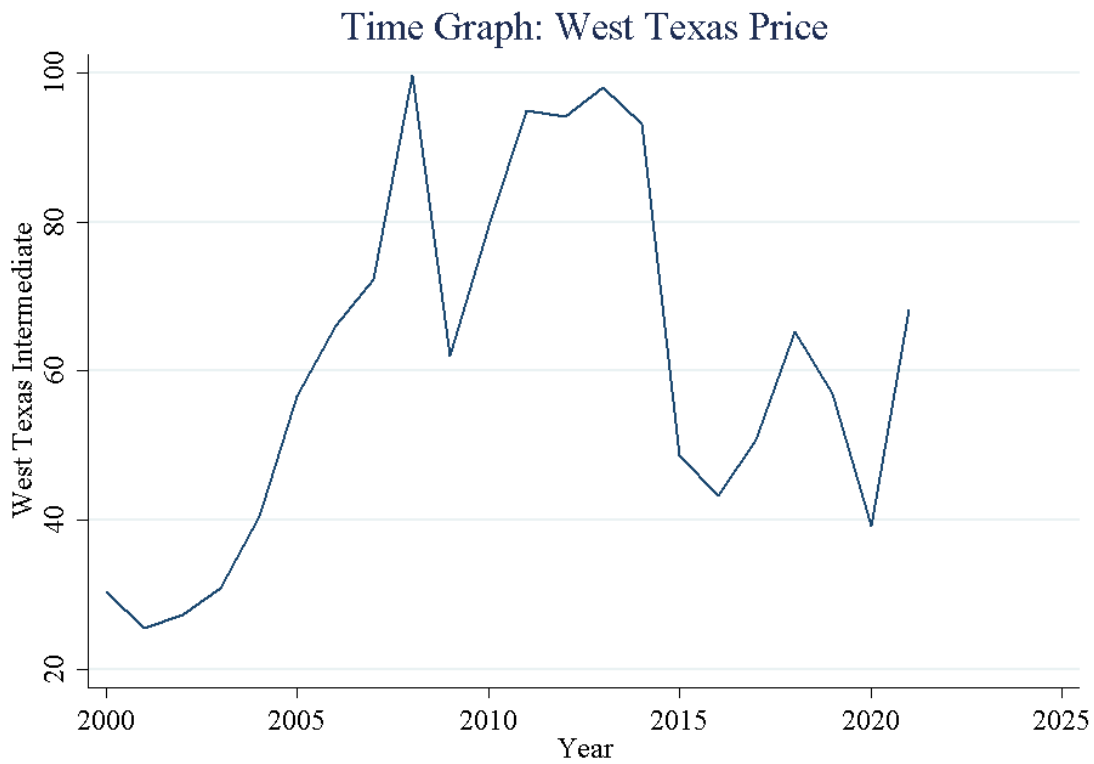
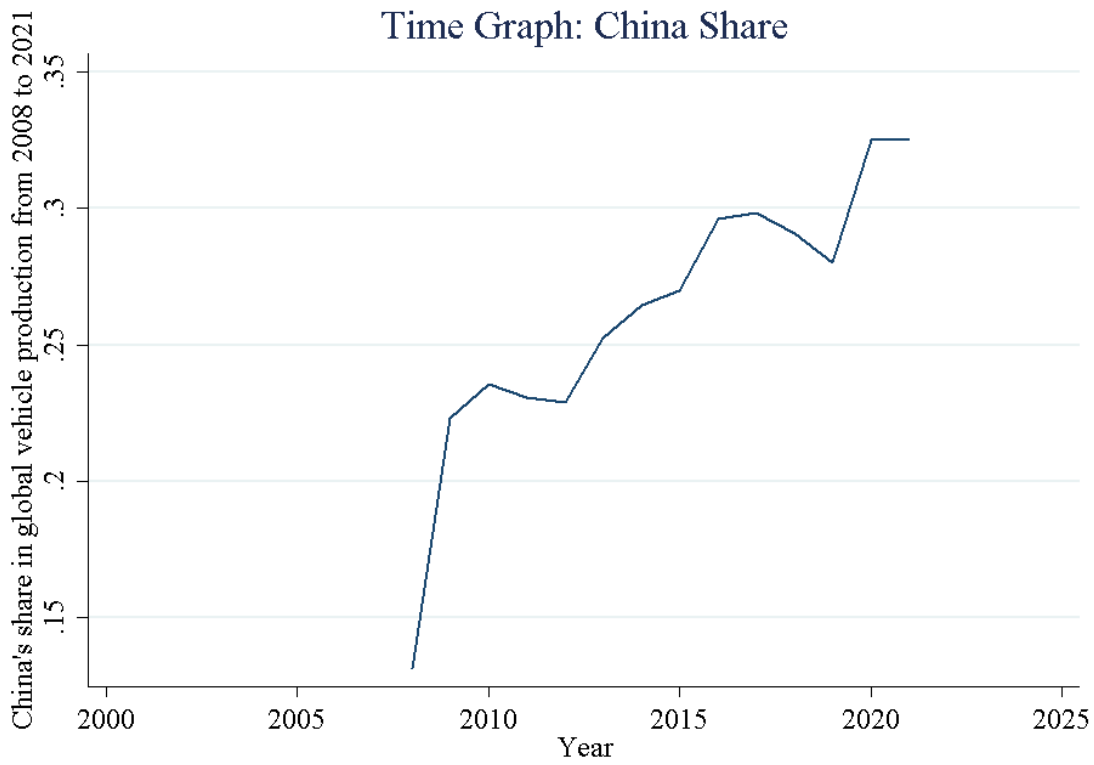
- Bauer, H. et al., 2020. How the semiconductor industry can emerge stronger after the COVID-19 crisis. *Advanced Electronics Practice*, pp. 1-8.
- Boquet, M., 2021. Key success factors in the semiconductor industry: Comparative analysis of the US, European and Chinese semiconductor industries. *Louvain School of Management, Université catholique*, pp. 1-91.
- Chun, H., Kim, H. & Roh, T., 2021. Supply Chain Ecosystem of Automotive Chip. *Electronics and Telecommunications Trends*, pp. 1-11.
- Hill, C. W. L., Schilling, M. A. & Jones, G. R., 2020. *Strategic Management: An Integrated Approach: Theory and Cases*. 13th ed. Boston: Cengage Learning, Inc..
- Jiang, Y., Shu, J. & Song, M., 2021. Coping with shortages caused by disruptive events in automobile supply chains. *Naval Research Logistics*, pp. 1-15.
- Johansson, E. & Westmark, K., 2021. Handling Volatility in Delivery Plans by Implementing Freeze Times - A Case Study of a Supplier in the Automotive Industry. *Chalmers University of Technology, Report No. E2021:032*, pp. 1-83.
- Llopis-Albert, C., Rubio, F. & Valero, F., 2021. Impact of digital transformation on the automotive industry. *Technological Forecasting & Social Change*, 162(120343), pp. 1-9.
- Marinova, G. I. & Bitri, A. K., 2021. Challenges and opportunities for semiconductor and electronic design automation industry in post-Covid-19 years. *IOP Conference Series: Materials Science and Engineering*, 1208(012036), pp. 1-6.
- Min, C. & Jianwen, L., 2020. Influence of COVID-19 on Manufacturing Industry and Corresponding Countermeasures from Supply Chain Perspective. *Journal of Shanghai Jiaotong University (Science)*, 25(4), pp. 409-416.

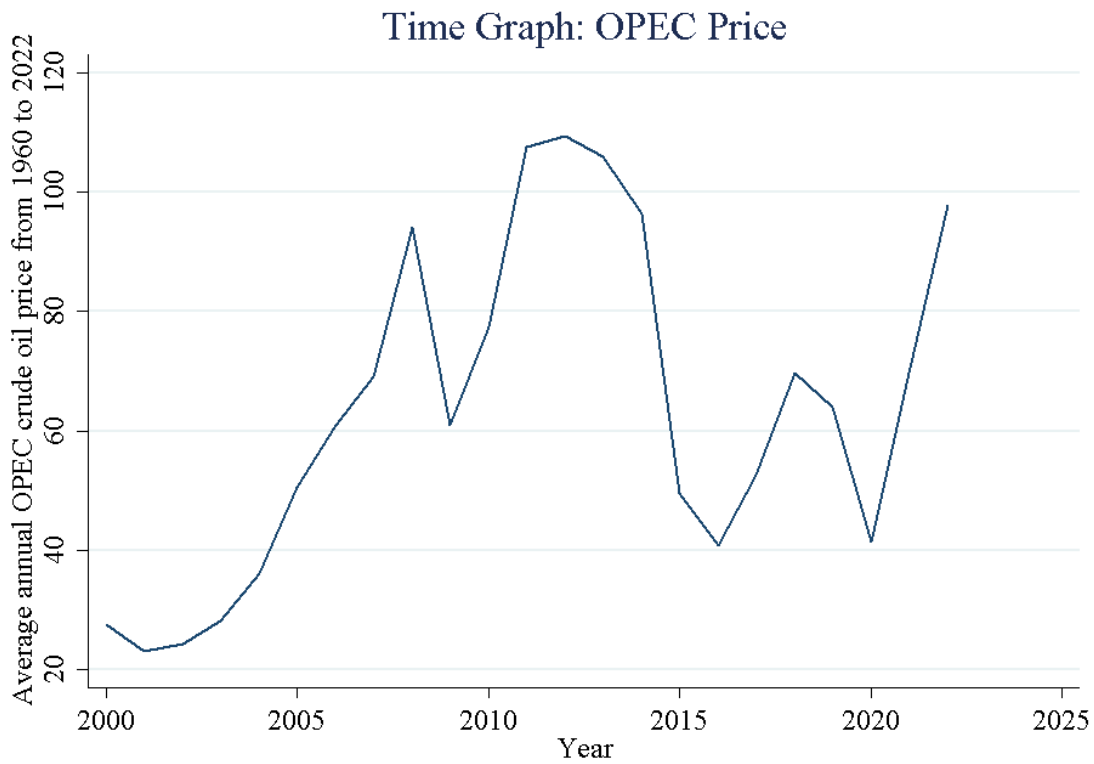
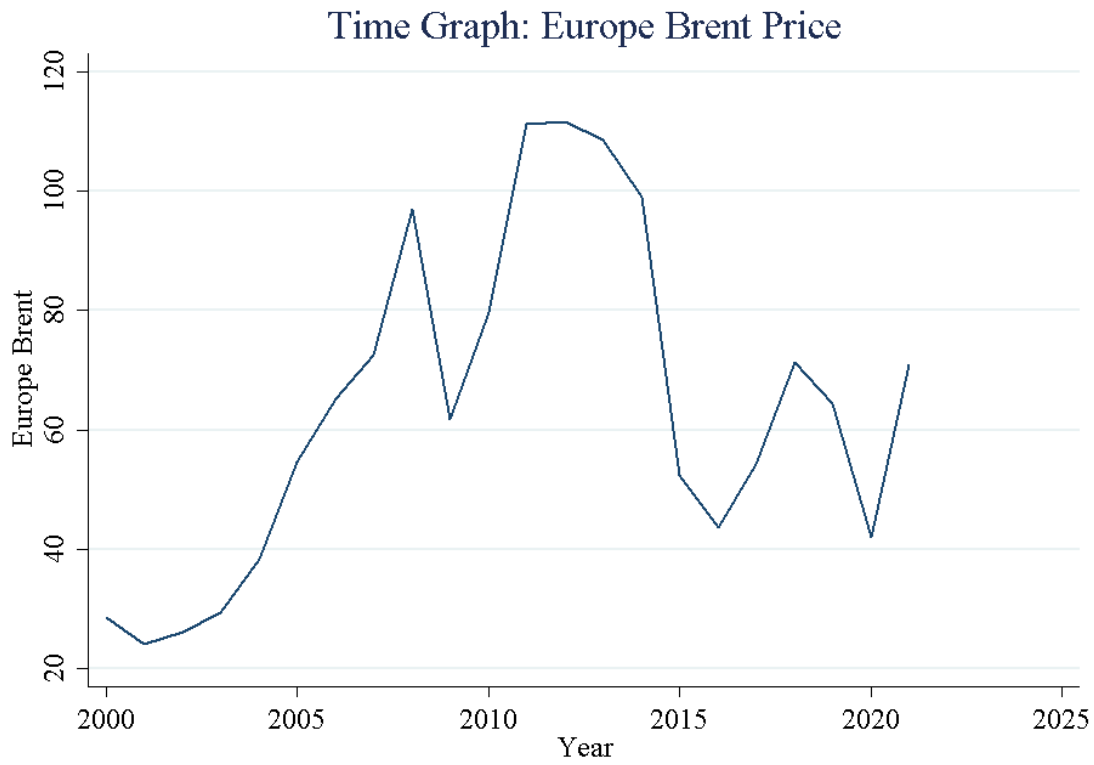
- Nayak, J. et al., 2021. An impact study of COVID-19 on six different industries: Automobile, energy and power, agriculture, education, travel and tourism and consumer electronics. *Expert Systems*, pp. 1-32.
- Schwab, K., 2021. *The Fourth Industrial Revolution*, s.l.: Encyclopedia Britannica.
- Statista, 2022. Capital expenditure in the global semiconductor industry from 2000 to 2021 (in billion U.S. dollars), s.l.: Thomas Alsop.
- Statista, 2022. China's share in global vehicle production from 2008 to 2021, s.l.: Statista Research Department.
- Statista, 2022. Electrified and battery electric vehicle sales as a percentage of car sales worldwide between 2020 and 2025, s.l.: Mathilde Carlier.
- Statista, 2022. Estimated passenger car production in selected countries in 2021 (in million units), s.l.: Statista Research Department.
- Statista, 2022. Estimated worldwide motor vehicle production from 2000 to 2021 (in million vehicles), s.l.: Martin Placek.
- Statista, 2022. Motor vehicle production of the United States and worldwide from 1999 to 2021, s.l.: Martin Placek.
- Statista, 2022. Projected revenue and production loss of the global automotive industry as a result of the semiconductor chip shortage in 2021(in billion U.S. dollars and million units), s.l.: Martin Placek.
- Statista, 2022. Semiconductor industry capital expenditure annual growth rate worldwide from 2000 to 2021, s.l.: Thomas Alsop.
- Statista, 2022. Semiconductor market size worldwide from 1987 to 2022 (in billion U.S. dollars), s.l.: Thomas Alsop.
- Statista, 2022. Semiconductor sales in China from 2015 to 2021, by month (in billion U.S. dollars), s.l.: Thomas Alsop.
- Statista, 2022. Semiconductor sales in the Americas from 2012 to 2021, by month (in billion U.S. dollars), s.l.: Thomas Alsop.
- Taiwan Semiconductor Manufacturing Company Limited, 2022. *A Look at Semiconductor Supply Chains*, s.l.: TSMC.
- Wooldridge, J. M., 2013. *Introductory Econometrics: A Modern Approach*. 5th ed. Mason: South-Western, Cengage Learning.
- World Semiconductor Trade Statistics, 2021. *WSTS Semiconductor Market Forecast Spring 2021*, San Jose: WSTS.
- Wu, X., Zhang, C. & Du, W., 2021. An Analysis on the Crisis of “Chips shortage” in Automobile Industry —Based on the Double Influence of COVID-19 and Trade Friction. *Journal of Physics: Conference Series*, 1971(012100), pp. 1-6.

## 9. APPENDICES

*Appendix A – Timeline Graphs*







**Appendix B – Analyses in Stata**

```
. ren Semiconductorunitshipmentswor semicondshipment
. ren Worldwidemotorvehiclesalesfr worldwidesale
. ren Numberofcarssoldworldwidebe worldwidedecar
. ren Chinasshareinglobalvehicle chinashare
. ren WestTexasIntermediate westtexas
. ren EuropeBrent europebrent
. ren AverageannualOPECcrudeoilpr annualopec
. ren Electrifiedvehiclesalesasap electricvehicle
. ren Batteryelectricvehiclesalesa battery
. ren Sizeofthememorycardmarketi sizememory
.
. tsset Year
      time variable: Year, 2000 to 2025
      delta: 1 unit

. tsline semicondshipment, graphregion(color(white)) bgcolor(white) title("Time Graph: Semicon
> ductor Shipment")

. graph export "semicondshipment.png", as(png) replace
(file semicondshipment.png written in PNG format)

. tsline worldwidesale, graphregion(color(white)) bgcolor(white) title("Time Graph: Worldwide
> Sale")

. graph export "worldwidesale.png", as(png) replace
(file worldwidesale.png written in PNG format)

. tsline chinashare, graphregion(color(white)) bgcolor(white) title("Time Graph: China Share")

. graph export "chinashare.png", as(png) replace
(file chinashare.png written in PNG format)

. tsline westtexas, graphregion(color(white)) bgcolor(white) title("Time Graph: West Texas Pri
> ce")

. graph export "westtexas.png", as(png) replace
(file westtexas.png written in PNG format)
```



```
. tsline europebrent, graphregion(color(white)) bgcolor(white) title("Time Graph: Europe Brent
> Price")

. graph export "europebrent.png", as(png) replace
(file europebrent.png written in PNG format)

. tsline annualopec, graphregion(color(white)) bgcolor(white) title("Time Graph: OPEC Price")

. graph export "annualopec.png", as(png) replace
(file annualopec.png written in PNG format)

.
. foreach i of varlist semicondshipment worldwidesale chinashare westtexas europebrent annualo
> pec {
  2. gen d`i'=D2.`i'
  3. }
(2 missing values generated)
(11 missing values generated)
(14 missing values generated)
(6 missing values generated)
(6 missing values generated)
(5 missing values generated)
. foreach i of varlist semicondshipment worldwidesale chinashare westtexas europebrent annualo
> pec {
  2. reg `i'
  3. dfuller `i', regress lags(3) trend
  4. estadd scalar DF=r(Zt)
  5. estadd scalar pvalue=r(p)
  6. eststo `i'
  7. }
```

Source	SS	df	MS	Number of obs	=	26
Model	0	0	.	F(0, 25)	=	0.00
Residual	2127020.85	25	85080.834	Prob > F	=	.
Total	2127020.85	25	85080.834	R-squared	=	0.0000
				Adj R-squared	=	0.0000
				Root MSE	=	291.69

semicondsh~t	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
_cons	780.3038	57.20437	13.64	0.000	662.4892 898.1184

Augmented Dickey-Fuller test for unit root                      Number of obs =                      22

Test Statistic	Interpolated Dickey-Fuller		
	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-3.293	-4.380	-3.600

MacKinnon approximate p-value for Z(t) = 0.0674

D.semicond~t	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
semicondsh~t					
L1.	-1.812613	.5504483	-3.29	0.005	-2.979511 -.6457149
LD.	1.044217	.4522404	2.31	0.035	.0855105 2.002924
L2D.	.2687669	.3155156	0.85	0.407	-.4000963 .93763
L3D.	.3868799	.2360932	1.64	0.121	-.1136153 .8873751
_trend	68.08262	20.49903	3.32	0.004	24.62662 111.5386
_cons	461.665	121.9024	3.79	0.002	203.2434 720.0866

```
added scalar:
      e(DF) = -3.2929763

added scalar:
      e(pvalue) = .06741521
```

Source	SS	df	MS	Number of obs	=	17
Model	0	0	.	F(0, 16)	=	0.00
Residual	1761.87981	16	110.117488	Prob > F	=	.
Total	1761.87981	16	110.117488	R-squared	=	0.0000
				Adj R-squared	=	0.0000
				Root MSE	=	10.494

worldwides~e	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
_cons	80.93412	2.545093	31.80	0.000	75.53876 86.32947

Augmented Dickey-Fuller test for unit root Number of obs = 13

Test Statistic	Interpolated Dickey-Fuller			
	1% Critical Value	5% Critical Value	10% Critical Value	
Z(t)	-1.062	-4.380	-3.600	-3.240

MacKinnon approximate p-value for Z(t) = 0.9351

D.worldwid~e	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
L1.	-.8212535	.7730327	-1.06	0.323	-2.649185 1.006678
LD.	.8170463	.8240683	0.99	0.354	-1.131566 2.765658
L2D.	.3064345	.6959044	0.44	0.673	-1.339118 1.951987
L3D.	.4327635	.6485147	0.67	0.526	-1.10073 1.966257
_trend	1.487212	2.056808	0.72	0.493	-3.376367 6.350791
_cons	52.96277	42.71685	1.24	0.255	-48.04653 153.9721

added scalar: e(DF) = -1.0623788

added scalar: e(pvalue) = .93511329

Source	SS	df	MS	Number of obs	=	14
Model	0	0	.	F(0, 13)	=	0.00
Residual	.033069732	13	.002543826	Prob > F	=	.
Total	.033069732	13	.002543826	R-squared	=	0.0000
				Adj R-squared	=	0.0000
				Root MSE	=	.05044

chinashare	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
_cons	.2608643	.0134797	19.35	0.000	.2317432 .2899854

Augmented Dickey-Fuller test for unit root Number of obs = 10

Test Statistic	Interpolated Dickey-Fuller			
	1% Critical Value	5% Critical Value	10% Critical Value	
Z(t)	-1.347	-4.380	-3.600	-3.240

MacKinnon approximate p-value for Z(t) = 0.8758

D.chinashare	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
chinashare					
L1.	-1.641482	1.218641	-1.35	0.249	-5.024971 1.742007
LD.	.5340399	.8670321	0.62	0.571	-1.873227 2.941307
L2D.	.3121213	.8643899	0.36	0.736	-2.08781 2.712052
L3D.	-.134125	.2642782	-0.51	0.638	-.8678788 .5996288
_trend	.0141662	.0119428	1.19	0.301	-.0189924 .0473247
_cons	.3338401	.2209791	1.51	0.205	-.2796962 .9473764

added scalar:

$$e(DF) = -1.3469781$$

added scalar:

$$e(pvalue) = .87581332$$

Source	SS	df	MS	Number of obs	=	22
-----				F(0, 21)	=	0.00
Model	0	0	.	Prob > F	=	.
Residual	12427.0225	21	591.762974	R-squared	=	0.0000
-----				Adj R-squared	=	0.0000
Total	12427.0225	21	591.762974	Root MSE	=	24.326

westtexas	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
-----						
_cons	61.06136	5.186359	11.77	0.000	50.27574	71.84699
-----						

Augmented Dickey-Fuller test for unit root                      Number of obs = 18

Test Statistic	----- Interpolated Dickey-Fuller -----			
	1% Critical Value	5% Critical Value	10% Critical Value	
Z(t)	-1.812	-4.380	-3.600	-3.240

MacKinnon approximate p-value for Z(t) = 0.6990

D.westtexas	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
-----						
westtexas						
L1.	-.4591796	.2534134	-1.81	0.095	-1.01132	.0929609
LD.	-.0156929	.3223924	-0.05	0.962	-.7181257	.6867399
L2D.	-.0794006	.3028599	-0.26	0.798	-.7392757	.5804745
L3D.	.0930941	.2902092	0.32	0.754	-.5392174	.7254057
_trend	-.7315346	1.070761	-0.68	0.507	-3.064522	1.601453
_cons	41.5848	18.4135	2.26	0.043	1.465228	81.70438
-----						

added scalar:

$$e(DF) = -1.8119781$$

added scalar:

$$e(pvalue) = .69895414$$

Source	SS	df	MS	Number of obs	=	22
Model	0	0	.	F(0, 21)	=	0.00
Residual	16585.6753	21	789.794063	Prob > F	=	.
				R-squared	=	0.0000
				Adj R-squared	=	0.0000
Total	16585.6753	21	789.794063	Root MSE	=	28.103

europaebrent	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
_cons	63.90409	5.991638	10.67	0.000	51.4438	76.36438

Augmented Dickey-Fuller test for unit root                      Number of obs = 18

Test Statistic	----- Interpolated Dickey-Fuller -----			
	1% Critical Value	5% Critical Value	10% Critical Value	
Z(t)	-1.776	-4.380	-3.600	-3.240

MacKinnon approximate p-value for Z(t) = 0.7163

D.europaebrent	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
europaebrent						
L1.	-.4379821	.2466316	-1.78	0.101	-.9753463	.0993821
LD.	.1912073	.3217732	0.59	0.563	-.5098763	.8922909
L2D.	-.111556	.297013	-0.38	0.714	-.7586916	.5355796
L3D.	.138083	.2994321	0.46	0.653	-.5143234	.7904894
_trend	-.361065	1.178	-0.31	0.764	-2.927706	2.205576
_cons	37.14407	18.15464	2.05	0.063	-2.41149	76.69962

added scalar:

$$e(DF) = -1.7758553$$

added scalar:

$$e(pvalue) = .7162825$$

Source	SS	df	MS	Number of obs	=	23
Model	0	0	.	F(0, 22)	=	0.00
Residual	17278.7677	22	785.398533	Prob > F	=	.
Total	17278.7677	22	785.398533	R-squared	=	0.0000
				Adj R-squared	=	0.0000
				Root MSE	=	28.025

annualopec	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
_cons	63.33652	5.843609	10.84	0.000	51.21762	75.45542

Augmented Dickey-Fuller test for unit root                      Number of obs = 19

Test Statistic	----- Interpolated Dickey-Fuller -----			
	1% Critical Value	5% Critical Value	10% Critical Value	
Z(t)	-4.380	-3.600	-3.240	-1.929

MacKinnon approximate p-value for Z(t) = 0.6396

D.annualopec	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
annualopec						
L1.	-.4677879	.2425309	-1.93	0.076	-.9917441	.0561683
LD.	.3249598	.2853505	1.14	0.275	-.2915024	.941422
L2D.	-.1345369	.2936898	-0.46	0.654	-.7690152	.4999413
L3D.	.1600318	.2956251	0.54	0.597	-.4786274	.7986909
_trend	.1625888	1.022843	0.16	0.876	-2.04713	2.372308
_cons	32.20834	17.12185	1.88	0.083	-4.781165	69.19786

added scalar:

$$e(DF) = -1.9287765$$

added scalar:

$$e(pvalue) = .6396299$$

```
. esttab semiconshipment worldwidesale chinashare westtexas europebrent annualopec using dful
> ler.rtf, stat(DF pvalue) replace label compress
(output written to dfuller.rtf)
```

```
.
. foreach i of varlist dsemiconshipment dworldwidesale dchinashare dwesttexas deuropebrent da
> nnualopec {
2. reg `i'
3. dfuller `i', regress lags(3) trend
4. estadd scalar DF2=r(Zt)
5. estadd scalar pvalue=r(p)
6. eststo `i'
7. }
```

Source	SS	df	MS	Number of obs	=	24
Model	0	0	.	F(0, 23)	=	0.00
Residual	167931.998	23	7301.39121	Prob > F	=	.
				R-squared	=	0.0000
				Adj R-squared	=	0.0000
Total	167931.998	23	7301.39121	Root MSE	=	85.448

dsemiconds~t	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
_cons	4.291667	17.44204	0.25	0.808	-31.78993 40.37327

Augmented Dickey-Fuller test for unit root                      Number of obs = 20

Test Statistic	----- Interpolated Dickey-Fuller -----			
	1% Critical Value	5% Critical Value	10% Critical Value	
Z(t)	-4.170	-4.380	-3.600	-3.240

MacKinnon approximate p-value for Z(t) = 0.0050

D.dsemicon~t	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
dsemiconds~t					
L1.	-3.643039	.8735662	-4.17	0.001	-5.516652 -1.769425
LD.	1.839663	.7111327	2.59	0.022	.314435 3.364891
L2D.	.7865096	.4709783	1.67	0.117	-.2236384 1.796658
L3D.	.3250277	.2461881	1.32	0.208	-.2029932 .8530486
_trend	1.148196	2.643768	0.43	0.671	-4.522124 6.818515
_cons	-14.27728	38.85448	-0.37	0.719	-97.61185 69.05729

added scalar:

e(DF2) = -4.1703061

added scalar:

e(pvalue) = .00496176

Source	SS	df	MS	Number of obs	=	15
Model	0	0	.	F(0, 14)	=	0.00
Residual	604.046789	14	43.1461992	Prob > F	=	.
				R-squared	=	0.0000
				Adj R-squared	=	0.0000
Total	604.046789	14	43.1461992	Root MSE	=	6.5686

dworldwide~e	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
_cons	.0986667	1.695999	0.06	0.954	-3.53889 3.736223

Augmented Dickey-Fuller test for unit root                      Number of obs = 11

Test Statistic	----- Interpolated Dickey-Fuller -----		
	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-4.380	-3.600	-3.240

MacKinnon approximate p-value for Z(t) = 0.2009

D.dworldwi~e	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
dworldwide~e					
L1.	-5.803443	2.080553	-2.79	0.038	-11.15168 -.4552112
LD.	3.180532	1.603027	1.98	0.104	-.940179 7.301244
L2D.	1.683788	.9463142	1.78	0.135	-.7487906 4.116366
L3D.	.6551144	.4676603	1.40	0.220	-.5470447 1.857274
_trend	-1.904293	1.128177	-1.69	0.152	-4.804366 .995779
_cons	14.73287	9.660438	1.53	0.188	-10.10008 39.56581

added scalar:  
e(DF2) = -2.7893752

added scalar:  
e(pvalue) = .20087421

Source	SS	df	MS	Number of obs	=	12
Model	0	0	.	F(0, 11)	=	0.00
Residual	.012917229	11	.001174294	Prob > F	=	.
Total	.012917229	11	.001174294	R-squared	=	0.0000
				Adj R-squared	=	0.0000
				Root MSE	=	.03427

dchinashare	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
_cons	-.0076083	.0098923	-0.77	0.458	-.0293812 .0141645

Augmented Dickey-Fuller test for unit root Number of obs = 8

Test Statistic	----- Interpolated Dickey-Fuller -----			
	1% Critical Value	5% Critical Value	10% Critical Value	
Z(t)	-1.783	-4.380	-3.600	-3.240

MacKinnon approximate p-value for Z(t) = 0.7130

D.dchinash-e	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
dchinashare					
L1.	-3.925986	2.20222	-1.78	0.217	-13.40137 5.549401
LD.	1.93566	1.760497	1.10	0.386	-5.639149 9.510469
L2D.	.8443323	1.046124	0.81	0.504	-3.656774 5.345439
L3D.	-.051763	.5066622	-0.10	0.928	-2.231754 2.128228
_trend	-.0016263	.0066234	-0.25	0.829	-.0301244 .0268719
_cons	.0126332	.0524314	0.24	0.832	-.2129607 .2382272

added scalar:  
e(DF2) = -1.7827404

added scalar:  
e(pvalue) = .7130206

Source	SS	df	MS	Number of obs	=	20
Model	0	0	.	F(0, 19)	=	0.00
Residual	14333.2496	19	754.381556	Prob > F	=	.
Total	14333.2496	19	754.381556	R-squared	=	0.0000
				Adj R-squared	=	0.0000
				Root MSE	=	27.466

dwesttexas	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
_cons	1.6875	6.141586	0.27	0.786	-11.16699 14.54199

Augmented Dickey-Fuller test for unit root Number of obs = 16

Test Statistic	----- Interpolated Dickey-Fuller -----			
	1% Critical Value	5% Critical Value	10% Critical Value	
Z(t)	-3.154	-4.380	-3.600	-3.240

MacKinnon approximate p-value for Z(t) = 0.0939

D.dwesttexas	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
dwesttexas					
L1.	-3.574876	1.133359	-3.15	0.010	-6.100158 -1.049594
LD.	1.669162	.949965	1.76	0.109	-.4474918 3.785816
L2D.	.8337686	.6547579	1.27	0.232	-.6251229 2.29266
L3D.	.3143024	.3227664	0.97	0.353	-.404866 1.033471
_trend	.8192913	1.46746	0.56	0.589	-2.450413 4.088995
_cons	-10.13361	18.14389	-0.56	0.589	-50.56072 30.2935



added scalar:

e(DF2) = -3.1542296

added scalar:

e(pvalue) = .09391378

Source	SS	df	MS	Number of obs	=	20
-----				F(0, 19)	=	0.00
Model	0	0	.	Prob > F	=	.
Residual	14692.6978	19	773.299883	R-squared	=	0.0000
-----				Adj R-squared	=	0.0000
Total	14692.6978	19	773.299883	Root MSE	=	27.808

deuropebrent	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
-----						
_cons	1.678	6.218118	0.27	0.790	-11.33667	14.69267
-----						

Augmented Dickey-Fuller test for unit root                      Number of obs = 16

Test Statistic	----- Interpolated Dickey-Fuller -----			
	1% Critical Value	5% Critical Value	10% Critical Value	
Z(t)	-3.204	-4.380	-3.600	-3.240

MacKinnon approximate p-value for Z(t) = 0.0836

D.deuropeb~t	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
-----						
deuropebrent	-----					
L1.	-3.239539	1.01109	-3.20	0.009	-5.492389	-.9866891
LD.	1.571032	.8460105	1.86	0.093	-.3139973	3.456061
L2D.	.8093492	.5944271	1.36	0.203	-.5151169	2.133815
L3D.	.388685	.3216581	1.21	0.255	-.3280138	1.105384
_trend	.7984509	1.575957	0.51	0.623	-2.713	4.309902
_cons	-9.422354	19.46986	-0.48	0.639	-52.80391	33.9592
-----						

added scalar:

$$e(DF2) = -3.2040052$$

added scalar:

$$e(pvalue) = .08360657$$

Source	SS	df	MS	Number of obs	=	21
Model	0	0	.	F(0, 20)	=	0.00
Residual	13957.2033	20	697.860167	Prob > F	=	.
				R-squared	=	0.0000
				Adj R-squared	=	0.0000
Total	13957.2033	20	697.860167	Root MSE	=	26.417

dannualopec	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
_cons	1.545714	5.764671	0.27	0.791	-10.47918	13.57061

Augmented Dickey-Fuller test for unit root                      Number of obs = 17

Test Statistic	----- Interpolated Dickey-Fuller -----			
	1% Critical Value	5% Critical Value	10% Critical Value	
Z(t)	-3.381	-4.380	-3.600	-3.240

MacKinnon approximate p-value for Z(t) = 0.0540

D.dannualo~c	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
dannualopec						
L1.	-3.250495	.9614509	-3.38	0.006	-5.366634	-1.134356
LD.	1.606781	.7962669	2.02	0.069	-.1457911	3.359352
L2D.	.8501181	.5610392	1.52	0.158	-.384721	2.084957
L3D.	.3969155	.3083101	1.29	0.224	-.2816704	1.075501
_trend	.8696813	1.358152	0.64	0.535	-2.119592	3.858955
_cons	-9.726225	17.53813	-0.55	0.590	-48.32738	28.87493

added scalar:

$$e(DF2) = -3.380823$$

added scalar:

$$e(pvalue) = .05400332$$

```
. esttab dsemicondshipment dworldwidesale dchinashare dwesttexas deuropebrent dannualopec usin
> g dfuller2.rtf, stat(DF2 pvalue) replace label compress
(output written to dfuller2.rtf)
```

```
.
.
. /*Engle and Granger Cointegration Test */
. foreach i of varlist worldwidesale chinashare westtexas europebrent annualopec {
  2. regress `i' semicondshipment
  3. predict resid_`i', residuals
  4. tsline resid_`i', graphregion(color(white)) bgcolor(white) title("Timeline for Residuals"
> )
  5. graph export residualsvecml_`i'.png, replace
  6. dfuller resid_`i', regress trend lags(3)
  7. regress D.`i' dsemicondshipment L.resid_`i'
  8. eststo vecm_`i'
  9. }
```

Source	SS	df	MS	Number of obs	=	17
Model	1135.51959	1	1135.51959	F(1, 15)	=	27.19
Residual	626.360223	15	41.7573482	Prob > F	=	0.0001
				R-squared	=	0.6445
				Adj R-squared	=	0.6208
Total	1761.87981	16	110.117488	Root MSE	=	6.462

worldwidesale	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
semicondshipment	.0441904	.0084742	5.21	0.000	.0261281	.0622526
_cons	45.94289	6.89069	6.67	0.000	31.25573	60.63005

(9 missing values generated)

(file residualsvecml\_worldwidesale.png written in PNG format)

Augmented Dickey-Fuller test for unit root                      Number of obs    =        13

Test Statistic	----- Interpolated Dickey-Fuller -----		
	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-1.706	-4.380	-3.600

MacKinnon approximate p-value for Z(t) = 0.7484

D.resid_wor1~e	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
resid_wor1~e						
L1.	-1.025712	.6013949	-1.71	0.132	-2.447784	.3963613
LD.	.6544368	.6790633	0.96	0.367	-.9512927	2.260166
L2D.	1.198269	.5382757	2.23	0.061	-.0745502	2.471089
L3D.	1.759076	.7902948	2.23	0.061	-.109674	3.627826
_trend	.4213146	.6914285	0.61	0.562	-1.213654	2.056283
_cons	-4.217651	6.002536	-0.70	0.505	-18.41139	9.976092

Source	SS	df	MS	Number of obs	=	16
Model	27.8944288	2	13.9472144	F(2, 13)	=	0.55
Residual	330.740071	13	25.4415439	Prob > F	=	0.5908
				R-squared	=	0.0778
				Adj R-squared	=	-0.0641
Total	358.6345	15	23.9089667	Root MSE	=	5.044

D.worldwidesale	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
dsemiconshipment	.0109004	.0144187	0.76	0.463	-.0202492	.04205
resid_worldwidesale						
L1.	-.1445377	.2453277	-0.59	0.566	-.674536	.3854605
_cons	1.124899	1.280571	0.88	0.396	-1.641606	3.891404

Source	SS	df	MS	Number of obs	=	14
Model	.025545495	1	.025545495	F(1, 12)	=	40.74
Residual	.007524237	12	.00062702	Prob > F	=	0.0000
				R-squared	=	0.7725
				Adj R-squared	=	0.7535
Total	.033069732	13	.002543826	Root MSE	=	.02504

chinashare	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
semiconshipment	.0002523	.0000395	6.38	0.000	.0001662	.0003384
_cons	.0492341	.0338246	1.46	0.171	-.0244633	.1229315

(12 missing values generated)

(file residualsvecml\_chinashare.png written in PNG format)

Augmented Dickey-Fuller test for unit root                      Number of obs =                      10

Test Statistic	----- Interpolated Dickey-Fuller -----		
	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-1.034	-4.380	-3.600

MacKinnon approximate p-value for Z(t) = 0.9394

D.resid_chin~e	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
resid_chin~e						
L1.	-1.473124	1.424977	-1.03	0.360	-5.429496	2.483248
LD.	.5501374	.8841	0.62	0.567	-1.904518	3.004793
L2D.	.175209	1.008006	0.17	0.870	-2.623466	2.973884
L3D.	-.0823839	.3902871	-0.21	0.843	-1.165995	1.001227
_trend	-.0016681	.0026402	-0.63	0.562	-.0089986	.0056623
_cons	.0172081	.0236875	0.73	0.508	-.048559	.0829751

Source	SS	df	MS	Number of obs	=	13
Model	.00750025	2	.003750125	F(2, 10)	=	20.82
Residual	.001801027	10	.000180103	Prob > F	=	0.0003
Total	.009301277	12	.000775106	R-squared	=	0.8064
				Adj R-squared	=	0.7676
				Root MSE	=	.01342

D.chinashare	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
dsemicondshipment	.0002009	.0000451	4.45	0.001	.0001003	.0003015
resid_chinashare						
L1.	-.978889	.1657376	-5.91	0.000	-1.348175	-.6096027
_cons	.0125701	.0037814	3.32	0.008	.0041446	.0209956

Source	SS	df	MS	Number of obs	=	22
-----+				F(1, 20)	=	2.52
Model	1391.59146	1	1391.59146	Prob > F	=	0.1279
Residual	11035.431	20	551.77155	R-squared	=	0.1120
-----+				Adj R-squared	=	0.0676
Total	12427.0225	21	591.762974	Root MSE	=	23.49

westtexas	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
-----+						
semicondshipment	.0339289	.0213645	1.59	0.128	-.0106367	.0784945
_cons	37.27444	15.79334	2.36	0.029	4.330121	70.21876

(4 missing values generated)

(file residualsvecml\_westtexas.png written in PNG format)

Augmented Dickey-Fuller test for unit root                      Number of obs = 18

Test Statistic	----- Interpolated Dickey-Fuller -----			
	1% Critical Value	5% Critical Value	10% Critical Value	
Z(t)	-1.801	-4.380	-3.600	-3.240

MacKinnon approximate p-value for Z(t) = 0.7041

D.resid_west	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
-----+						
resid_west						
L1.	-.4398716	.2441925	-1.80	0.097	-.9719215	.0921782
LD.	-.029888	.3172613	-0.09	0.927	-.7211411	.6613651
L2D.	-.0453777	.3005033	-0.15	0.882	-.7001182	.6093628
L3D.	.1043129	.2889878	0.36	0.724	-.5253374	.7339632
_trend	-1.258156	.9935435	-1.27	0.229	-3.422901	.9065894
_cons	18.14804	13.23854	1.37	0.196	-10.69627	46.99234

Source	SS	df	MS	Number of obs	=	20
-----+				F(2, 17)	=	2.46
Model	1495.34417	2	747.672086	Prob > F	=	0.1151
Residual	5162.31708	17	303.665711	R-squared	=	0.2246
-----+				Adj R-squared	=	0.1334
Total	6657.66126	19	350.403224	Root MSE	=	17.426

D.westtexas	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
dsemicondshipment	-.0396743	.0492171	-0.81	0.431	-.1435133	.0641648
resid_westtexas						
L1.	-.385985	.1752679	-2.20	0.042	-.7557679	-.016202
_cons	3.061054	3.942183	0.78	0.448	-5.256226	11.37833

Source	SS	df	MS	Number of obs	=	22
Model	2544.64372	1	2544.64372	F(1, 20)	=	3.62
Residual	14041.0316	20	702.051581	Prob > F	=	0.0714
				R-squared	=	0.1534
				Adj R-squared	=	0.1111
Total	16585.6753	21	789.794063	Root MSE	=	26.496

europerebrent	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
semicondshipment	.0458804	.024099	1.90	0.071	-.0043891	.0961499
_cons	31.73818	17.81471	1.78	0.090	-5.422652	68.89901

(4 missing values generated)  
(file residualsvecml\_europerebrent.png written in PNG format)

Augmented Dickey-Fuller test for unit root                      Number of obs    =        18

Test Statistic	----- Interpolated Dickey-Fuller -----		
	1% Critical Value	5% Critical Value	10% Critical Value
Z(t)	-1.758	-4.380	-3.240

MacKinnon approximate p-value for Z(t) = 0.7248

D.resid_eu~t	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
resid_euro~t						
L1.	-.4163504	.236885	-1.76	0.104	-.9324785	.0997777
LD.	.1549153	.3159873	0.49	0.633	-.5335618	.8433924
L2D.	-.0667252	.2941857	-0.23	0.824	-.7077008	.5742505
L3D.	.1477733	.2977982	0.50	0.629	-.5010732	.7966197
_trend	-1.079355	1.05354	-1.02	0.326	-3.374823	1.216112
_cons	15.84599	14.00754	1.13	0.280	-14.67381	46.3658

Source	SS	df	MS	Number of obs	=	20
Model	1721.89213	2	860.946067	F(2, 17)	=	2.44
Residual	5995.90589	17	352.700346	Prob > F	=	0.1170
				R-squared	=	0.2231
				Adj R-squared	=	0.1317
Total	7717.79802	19	406.199896	Root MSE	=	18.78

D.europerebrent	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
dsemicondshipment	-.0546732	.0527813	-1.04	0.315	-.1660321	.0566856
resid_europerebrent						
L1.	-.3562099	.1668072	-2.14	0.048	-.7081424	-.0042774
_cons	3.49558	4.24951	0.82	0.422	-5.470102	12.46126

Source	SS	df	MS	Number of obs	=	23
-----				F(1, 21)	=	5.64
Model	3656.11573	1	3656.11573	Prob > F	=	0.0272
Residual	13622.652	21	648.697714	R-squared	=	0.2116
-----				Adj R-squared	=	0.1741
Total	17278.7677	22	785.398533	Root MSE	=	25.47

annualopec	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
-----					
semicondshipment	.0512536	.0215891	2.37	0.027	.0063565 .0961507
_cons	26.42904	16.42833	1.61	0.123	-7.735548 60.59363

(3 missing values generated)  
 (file residualsvecml\_annualopec.png written in PNG format)  
 Augmented Dickey-Fuller test for unit root                      Number of obs = 19

Test Statistic	----- Interpolated Dickey-Fuller -----			
	1% Critical Value	5% Critical Value	10% Critical Value	
Z(t)	-1.950	-4.380	-3.600	-3.240

MacKinnon approximate p-value for Z(t) = 0.6285

D.resid_anno~c	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
-----					
resid_anno~c					
L1.	-.4565147	.2341468	-1.95	0.073	-.962358 .0493287
LD.	.3026616	.2858779	1.06	0.309	-.31494 .9202632
L2D.	-.0869497	.2941893	-0.30	0.772	-.722507 .5486075
L3D.	.1905786	.2949033	0.65	0.529	-.4465211 .8276784
_trend	-.634939	.9315122	-0.68	0.507	-2.647349 1.377471
_cons	10.86368	12.98716	0.84	0.418	-17.19338 38.92073

Source	SS	df	MS	Number of obs	=	21
-----				F(2, 18)	=	3.21
Model	2113.59631	2	1056.79815	Prob > F	=	0.0643
Residual	5929.92175	18	329.440097	R-squared	=	0.2628
-----				Adj R-squared	=	0.1809
Total	8043.51806	20	402.175903	Root MSE	=	18.15

D.annualopec	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
-----					
dsemicondshipment	-.0649516	.0470025	-1.38	0.184	-.1637002 .0337971
resid_annualopec					
L1.	-.3686831	.1603375	-2.30	0.034	-.7055396 -.0318265
_cons	3.870648	3.964202	0.98	0.342	-4.457831 12.19913

```
. esttab vecm_worldwidesale vecm_chinashare vecm_westtexas vecm_europebrent vecm_annualopec //
> /
> using vecm.rtf, replace starlevels(* 0.1 ** 0.05 *** 0.01) se label
(output written to vecm.rtf)

.
. log close
  name: <unnamed>
  log: D:\Courses\Econ\vecm.log
  log type: text
  closed on: 1 Jun 2022, 01:27:43
```