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**First Year Master Thesis**  
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## **Is Chernobyl radiation the main cause of cancer deaths? The research on cancer mortality before and after 1986**

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### **Abstract**

The Chernobyl catastrophe is still referred to as the biggest radioactivity fallout problem in the history of the world. The cloud that was released from the broken reactor traveled across larger parts of the European continent and the after-effects of the radiation are said to be felt and faced throughout the impacted regions up to this day – and that this process is far from over.

This thesis aims at examining the size as well as the significance of the effects of cancer-related mortality on people after the catastrophe in 1986 compared to the cancer deaths that occurred before the Chernobyl accident. The time period that is used in this paper spans from 1966 up to 2005, which equals to twenty years before and twenty years after the reactor explosion. Within this period two countries are examined. Poland is used as the treatment group as that country is located very close to the danger zone, whereas Spain is used as a control group because it is located outside of the impacted area. Two cancer types are examined within both of these countries. Lung cancer is examined as the one caused by radiation fallout, while stomach cancer is used as the one with the development not impacted by the radioactive particles. Moreover, a number of other factors, besides radiation, is used to statistically establish whether or not the radiation is the sole and significant determinant of cancer. These are: smoking habits, obesity and air pollution.

The results show that the Chernobyl's radiation still affects lung cancer mortality of people in Poland, but to a much lesser degree than the polluted air or even heavy smoking. Stomach cancer in Poland is caused by a combination of all of the modeled variables. Spanish lung cancer mortality is also affected by heavy smoking and air pollution, whereas stomach cancer deaths in this country are caused mainly by excessive nicotine intake.

**Keywords:** Cancer; Chernobyl disaster; mortality; lung; stomach; radioactivity; fallout.

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# 1. Introduction

The catastrophe in Chernobyl came to be a known and popular topic in the media, politics and various discussion panels. Whenever a topic of radiation is concerned, the accident in the Ukrainian nuclear power plant is brought up as the biggest accident in terms of fallout and as the one with voluminous long-term consequences and after-effects. The radiation impacts the health and lives of flora and fauna as well as human beings in the contaminated regions.

As the radioactive cloud spread over a large part of Europe, the soil in those regions was impacted by the particles. In turn, many plants, including fruits and vegetables, were deemed inedible and new methods of growing and cleaning had to be established. In many regions it is of continuous danger to pick up berries and mushrooms. A fact that is worth mentioning is that the Ukrainian Worm Wood Forest, which was the name of the pine forest surrounding the area of the power plant and the nearby city of Pripyat, had to have its name changed to Red Forest as the trees turned their color to a dead brownish red after the catastrophe. The forest is located in the alienation zone, which is the area of 30-kilometer radius around the Chernobyl Nuclear Power Plant (CNPP). This zone is referred to as the area where the doses of radiation are still the biggest, most dangerous and quite active. Animals were also the victims of the radiation. It is still said that the radiation particles are present in the earth of many regions where the cloud flew over. Because of the contamination a number of fish and birds have been severely mutated or their development has been stunted. Many species of boar and deer may still be under the radioactive influence on many impacted territories. Similar methods, as in the case of plants, had to be created in order to make animal products safe for the market and people again. Many farms in the contaminated areas still, after all those years, have restrictions put on them in terms of which kinds of livestock (especially cows and sheep) that can be sent out onto the market.

Human beings were impacted by radiation as well. Many deaths, cancer types and chromosomal changes were registered in the years following the catastrophe in the reactor number four. The most prevalent types of cancers that are connected to the radiation fallout are thyroid and leukemia. Other common types include lung cancer as well as prostate, rectum and breast ones. Apart from cancers, a big number of other non-cancer diseases and problems were also introduced as being connected to the fallout influence. Among these, issues with the respiratory and digestive systems are worth mentioning. In addition, radiated people were often diagnosed with lymphatic illnesses and even abnormal aging. According to the Greenpeace report (2006), people living in the contaminated areas (especially in Belarus and Ukraine) can have their biological age exceeding the natural calendar one by up to nine years. There was also a magnitude of Down syndrome cases among children born in Poland, Germany and Austria between 1987 and 1989. “A statistically significant increase in frequency was apparent in January 1987, corresponding to children conceived during the period of maximal Chernobyl fallout.” (Greenpeace 2006, p. 131).

## *1.1. The story of the accident*

The Chernobyl disaster occurred at the Chernobyl Nuclear Power Plant in Ukraine. “The plant was located approximately 100 km north of the Ukrainian capital Kiev, near the border with Belarus.” (Rosen 2006, p. 1). On 26<sup>th</sup> April 1986, Saturday, around 1:23 at night, reactor number four suffered a dangerous power increase, which led to the explosion in the reactor’s core. When the reactor’s graphite moderator was exposed to the atmosphere it went aflame.

The resulting smoke turned into a nuclear cloud. At the same time the reactor itself emitted large quantities of fuel and core materials into the air. Since the reactor was not encased in any type of protection, the pollution was able to spread quite easily. It is a known fact that the whole accident took place, ironically, during a scheduled test of the reactor's emergency core cooling unit.

The contamination spread over large parts of the European continent in a short matter of time. It is said that the cloud traveled westward as far as Austria and Switzerland, northward it reached Scandinavia, while in the south it spread to Italy and Greece. Clark (1986) wrote that the deposition of radionuclides took place in two ways during the passage of the cloud: dry and wet; the latter one is connected with rainfall. "Radioactive fallouts from the Chernobyl clouds touched many territories, where more than three billion people live." (Greenpeace 2006, p. 8). Then it is added in that report that more than half of the territories in thirteen nations were contaminated by the radiation from the power plant and about one-third was in eight additional nations. Most of Eastern European countries experienced several days of very clear and sunny weather during the several days following that 26<sup>th</sup> April. Only later, it became known to people in these nations that such cloudless sky was caused by the fact that the radiation cloud was passing by.

**Figure 1.1.1** Map of the area majorly contaminated by the Chernobyl cloud  
Source: [http://www.our-energy.com/chernobyl\\_disaster.html](http://www.our-energy.com/chernobyl_disaster.html) (accessed: 04.05.2011)



There was no information and no media coverage on the subject from the Soviet Union's side until the third day after the catastrophe, so the effects of the fallout were already taking place way before any sort of preventive methods were put into action. Soviet Union tried to hush this event up and was succeeding until the alarms of high radiation levels were set off in the Swedish power plant located in Forsmark (with the distance of over one thousand kilometers from Chernobyl). Only then Soviet authorities began to spread the news about the accident along with the hints that by the 29<sup>th</sup> April emissions were far from over. The preventive actions in all of the impacted nations were mostly about giving people doses of iodine, so as to protect their bodies and organisms from absorbing the contaminated iodine from the radiation. The most common version of this iodine is called the Lugol's solution, which is a mixture elemental iodine and potassium iodide in water. Polish people were advised by their

government, contrary to what the Soviet authorities were ordering, to drink that solution for protection against the radioactive iodine in the first days after the catastrophe<sup>1</sup>.

The catastrophe had the biggest effect, of course, on those that were in close proximity to the broken reactor. Workers who were present in the plant during the explosion died instantly. Many of the firefighters brought in to extinguish the flames that were injured during this action died while being hospitalized. Some of them died from large and acute doses of radiation. The inhabitants of Pripyat were ordered to evacuate the city as late as the second day after the disaster. About 100000 people had to be relocated behind the borders of the danger zone. There is also a known case of the “liquidators”, the circa 300 000 up to 600 000 clean-up workers that were recruited from the Soviet and Baltic nations to dispose of the debris and to clean the overall alienation area. A lot of workers were experiencing major health problems after the work was done, which was the result of the radiation doses received by their bodies despite wearing the protective gear. This was especially true for those that entered in close contact with the power plant. In some cases, even the offspring of these people had some physical mutations. In summary, there were about 31 immediate casualties: 28 from maximum exposure to radiation, two from the explosion and one from a heart attack. In addition, 238 were diagnosed with injuries caused by acute radiation. No one can tell for sure, though, how many additional radiation victims were and still are out there on the contaminated territories. “In addition to the lack of reliable information provided to the people affected in the first few years after the accident, there was widespread mistrust of official information and the false attribution of most health problems to radiation exposure from Chernobyl.” (World Health Organization 2006, Fact sheet N° 303).

Nonetheless, the effects of the catastrophe on human health, because of the traveling cloud, are said to be visible up to the present times and in a quite big area. It also seems very plausible to imply that the after-effects of radiation will continue to influence us in the upcoming future. “(...) it is reasonable to conclude that the Chernobyl accident has caused, and will continue to cause, a significant amount of morbidity and mortality across Europe (...) and beyond.” (Greenpeace 2006, p. 135). The aim of this thesis is to examine how big and significant were and are the effects of cancer mortality on people after the catastrophe compared to the cancer deaths that occurred before the Chernobyl accident.

## *1.2. Problems and discrepancies*

My initial idea was to write about and research thyroid and leukemia cancers since those are the most prominent types caused by radiation. The data for leukemia was complete in terms of observations, but had to be put aside as there are no major and known deterministic factors other than radiation that are causing this cancer type. On the other hand, the data for thyroid cancer mortality in Poland was quite incomplete. This resulted in more than twenty years worth of yearly observations that were missing from my dataset.

I started to pursue this problem wanting to find the missing thyroid observation and as the result I found out that it might have been a combined effect of the government’s cover ups and strikes in the medical industry that stood for the greater part of missing observations. The strikes of the medical staff took place in the late 1990s. Cover ups about the death toll, radiation spread and its consequences were done during the Soviet times and especially in the

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<sup>1</sup> This fact is given here courtesy of my parents.

post-Chernobyl days. These resulted in missing or misleading information. This is why one may get varying estimates of the mortality figures depending on the sources. This is quite apparent in the above paragraph in which I mention the Pripyat evacuation, “liquidators” and figures on deaths – one can never be completely sure about these numbers.

There are two sides to the nuclear power plant argument. This also adds to the confusion in terms of data and information. Scientists are apt to say that even though the effects of Chernobyl were big and grave at the time, it is mostly in the past by now. They may also add that nuclear energy is still in big demand. Environmental agencies, on the contrary, will state that the health consequences are very far from being over and that all of the power plants should be shut down and closed. This is another aim of my thesis: to see which of the two sides of this argument is more correct in its assumptions about the after-effects of the Chernobyl disaster.

## 2. Data

The time period that is examined in this paper spans between 1966 and 2005, which makes up to twenty years before and twenty years after the catastrophe. Since the accident in Chernobyl occurred quite early in the first half of the year 1986, so for simplicity reasons this year is placed in the “after” period. I am using a dummy variable for indication of these two separate periods. Poland is researched as the treatment group as it is located in the contaminated area as well as quite close to the power plant (CNPP was positioned about 450 kilometers from the eastern Polish border). Spain, on the other hand, is used as a control group as this country was not impacted by the radiation from the Chernobyl cloud.

In both nations two genders are examined along with mortality because of two types of cancer per each gender. These chosen cancer types are stomach cancer and lung cancer. Lung cancer is researched as it is a type that is clearly caused by radiation. Stomach cancer can also be caused by radiation, but to a much lesser degree; it is, thus, used for comparison purposes. “The total predicted number of cases possibly attributable to Chernobyl in Europe (whose population were more 570 million people in 1986) up to 2065 is large in absolute terms, about 23 000 for all cancers excluding leukemia, thyroid cancer and nonmelanoma skin cancer (...).” (Cardis et al. 2006, p. 1230). In order to determine whether or not the radiation is a sole and major factor contributing to cancer mortality, additional variables are furthermore used for both countries. Overweight people are more probable to be the victims of the stomach cancer – obesity can be, thus, treated as a deterministic factor of that cancer type. People living in the areas with high air pollution are running a higher risk of being diagnosed with lung cancer – air pollution can, as such, be treated as the deterministic factor of that type of cancer. In addition, both chosen cancer types are said to be caused by excessive smoking and nicotine intake<sup>2</sup>. It has to be noted that general mortality numbers for both genders in both countries are also, of course, incorporated into the analysis.

In addition, age-specific mortality rates for each cancer type per each gender in each of the two nations are included. This is done in order to get a better and more specific outlook on the differences between the two periods. There are only three age groups that are used: 0-29 years, 30-59 years and 60-85+. The latter group may seem to be the broadest, but as the age progresses the number of observations gets rapidly smaller. For people at the age of 90+, mortality in both countries decreases to very small single-digit figures. Why only three age groups? It is done so for the purpose of using chi-square cross-tabulations testing method; more about this is explained in the Methods section.

### 2.1. Dataset sources

General mortality rates for Poland and Spain are taken from website that belongs to Human Mortality Database (HMD). All of the data for cancer mortality along with age-specific observations is taken from the International Agency for Research on Cancer (IARC) website. IARC is a part of the World Health Organization (WHO). Data on the cancer-determining variables consists of: number of smoked cigarettes in million pieces per year (proxy for

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<sup>2</sup> Information about the deterministic causes of cancers is taken from the online articles located on the MedicineNet website (<http://www.medicinenet.com>) as well as from the book entitled “Onkologi” (2008) written by Dalianis, Henriksson and Ringborg. For further information, please check the “Sources” section at the end of this paper.

smoking habits), available fat per person per day in grams (proxy for obesity) as well as per capita emissions of sulfur dioxide, SO<sub>2</sub>, in kilograms per year (proxy for air pollution). This whole part of the dataset also comes from the WHO.<sup>3</sup>

There are certain drawbacks with the dataset:

- a) It needs to be noted that there are missing observations for Poland for both cancer types as well as for smoking-related deaths in 1997 and 1998. This is because of the aforementioned worker strikes among the medical staff.
- b) The part of the dataset that is describing the smoking habits does not cover the whole examined period of time – it covers the years between 1970 and 2000 (for both countries).
- c) The part of the dataset that is describing the obesity does not cover the whole examined period of time – it covers the period between 1970 and 2005 (for both countries).
- d) The part of the dataset that is describing the air pollution does not cover the whole examined period of time – it covers the time between 1970 and 2000 (for both countries).
- e) The problem with all of the deterministic variables is that they are not gender-specific.

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<sup>3</sup> For the information on all of the used variables along with their names and their descriptions, please refer to tables A.2.1 and A.2.2, both of which are located in the Appendix section of this thesis.

### 3. Methods

In order to examine the data two kinds of tests are performed. The first uses the cross-tabulation method with Pearson Chi-square tests. The second one is done with the help of the time series analysis methods. The reason for using two tests instead of one is to get a better and deeper look at the differences between the two time periods – before and after the Chernobyl catastrophe in terms of cancer mortality. The first method is performed by using the aforementioned age-specific cancer mortality figures. It is done with the help of the statistical software called SPSS. The second test is done by creating statistical models with the help of the general cancer mortality numbers and the bespoke deterministic variables. Program entitled STATA is used in this part. Below follows information, rules and restrictions of those two chosen test types as well as author's theory-to-practice approaches.

#### 3.1. Cross-tabulations

Analysis of charts created by cross-tabulations is done by using the Pearson Chi-square ( $\chi^2$ ) method. This method is based on comparing the observed frequencies in a table with their expected counterpart values. The formula for this test is located below (formula (1)). In this equation O stands for observed values, E is for estimated frequencies, while subscripts r and c stand for rows and respectively columns.

$$\chi^2 = \sum_{r=1}^{\infty} \sum_{c=1}^{\infty} \frac{(O_{rc} - E_{rc})^2}{E_{rc}} \quad (1)$$

“It was proposed in 1900 by Karl Pearson, the British statistician (...).” (Agresti 2007, p. 35). The test is done by creating a difference between the observed and awaited values in each cell of the contingency table. This difference is then taken to its square power and divided by the estimated frequencies. The test statistic, as such, is the sum of all these ratios. It takes the value of zero only when  $O_{rc} = E_{rc}$ . The  $\chi^2$ -test can never attain negative value as the difference is squared. Bigger differences between the observed and expected values create greater values of chi-square. This, in turn, means a stronger support for the alternative hypothesis.

The null hypothesis is rejected if the calculated sum is bigger than the critical value for a given significance level. The amount of the degrees of freedom (df) is calculated by using the formula (2):

$$df = (r - 1) \times (c - 1) \quad (2)$$

Under  $H_0$  there are  $r - 1$  unique row probabilities and  $c - 1$  sole column probabilities. This means that there are  $(r - 1) + (c - 1)$  parameters. Since the probability cannot exceed one, so these parameters can be equal, at the maximum, to  $r \times c - 1$ . The whole derivation is placed in formula (3) below:

$$\begin{aligned} df &= (r \times c - 1) - [(r - 1) + (c - 1)] = r \times c - 1 - r + 1 - c + 1 = r \times c + 1 - r - c = \\ &= (r - 1) \times (c - 1) \end{aligned} \quad (3)$$

It is the null hypothesis that decides the structure and distribution of the observations. The main idea is that there is independence between the time periods and each of the tested variables. This means that it is assumed that the number of observations for each variable is equal in both time periods.

The significance of the p-values is defined in a standard statistical manner: one-star  $10\% \leq P < 5\%$ , two-star  $1\% < P \leq 5\%$  and three-star  $P \leq 1\%$ . The latter one is when the significance is the strongest and when the null hypothesis is most likely to be rejected. In this thesis a significance level of  $\alpha = 5\%$  is used.

For the test to work all table cells need to have expected values that are bigger than one. In addition only 20% of all the cells can have estimations below the value of five. Agresti (2007) points out that this test, like any other significance test, has limitations. It indicates whether there is an association between the variables, but it does not give any information about its power or direction. “Rather than relying solely on these tests, study the nature of the association. It is sensible to study residuals (...).” (Agresti 2007, p. 40).

Standardized residuals are, thus, being used in the contingency tables as they are providing a deeper and more detailed look into the nature of the relationship. According to Agresti (2007) there is a lack of fit of the null hypothesis in the table cell when the residual in that very cell exceeds the value of two (when there are few cells in the table) and of three (when there are many cells). It has to be noted that in the table with equal amount of columns and rows the variable residuals will have the same absolute values (one will be positive, the other one negative) between the periods. “This is because the observed counts and the estimated expected frequencies have the same row and column totals.” (Agresti 2007, p. 39).

All in all, Pearson chi-square cross-tabulations are usually used as a basic background test that gives a bit more scientific representation of the data. This testing method is used in order to show the differences between the two examined time periods. In addition, it allows one to observe how much the results differ from their expected values. Adding residuals to the contingency tables helps in realizing how the results are placed in comparison to the accepted levels for the null hypothesis of equal amounts of observations in both time periods.

### 3.2. Time series analysis

Time series is the name for the variables that have their values spread across as well as being dependent on time. The most common model in time series is called AR(1) or an autoregressive process of order one; this model is used in this part of the paper to describe time series. This is the most basic type of the AR(p) models. AR(1) can be written as shown in formula (1) below:

$$y_t = \rho \times y_{t-1} + u_t \quad (1)$$

In that equation t is the time index (1,...,T),  $u_t$  is the so-called white noise with a zero-mean and a variance that is equal to a constant  $\sigma^2$ . The parameter  $\rho$  is called as an autoregressive coefficient and it has an absolute value equal to less than one. In this equation the dependent variable  $y_t$  is determined by its own past effect ( $y_{t-1}$ , also known as the first lag), with a variable impact that is equal to  $\rho$ . It is also determined by a random shock occurring at time t, the aforementioned white noise factor. This equation can be extended by adding certain

components, e.g. a constant and a trend. In addition, more difficult models can be made by using autoregressive processes of higher orders than 1 (up to AR(p)), as well as by adding moving average processes of various orders (MA(q)) to the equation. It needs to be mentioned that all of the information and methods mentioned in this section can be also applied to more difficult models and multivariate cases.

Time series can be stationary or non-stationary (unit roots). The difference is dependent on whether or not certain assumptions are violated. In stationary series the observations are fluctuating around the constant mean. Stationary series, moreover, need to have a constant variance. The covariance between two random observations within these series should not time-dependent. This means that for stationarity all of the below equations need to be true (formula (2)):

$$\begin{aligned} E(y_t) &= \mu \\ \text{Var}(y_t) &= \sigma^2 \\ \text{Cov}(y_t, y_{t+s}) &= \text{Cov}(y_t, y_{t-s}) = \gamma_s \end{aligned} \quad (2)$$

It is quite important to differentiate between those two types of series as “(...) there is a danger of obtaining apparently significant regression results from unrelated data when non-stationary series are used in regression analysis.” (Hill et al. 2008, p. 333). This is called a spurious regression. One has to use the Augmented Dickey-Fuller test (ADF) to test the type of the series with a regard to the inclusion or exclusion of constant and a trend. Dickey-Fuller tests have their own critical values that are called  $\tau$ -statistics (tau). They tend to be more negative than standard t-statistics; therefore the null hypothesis of unit root is less likely to be rejected. In addition, these critical values change depending on the chosen deterministic components of the equation. The number of lags is decided upon by examining an ACF chart (autocorrelation function). The null hypothesis, which stands here for the presence of unit roots in the series, is rejected when the calculated tau is less than the critical one.

To make a unit root series into a stationary one it has to be differenced. There is a general rule that a series integrated of order d (I(d)) needs to be differenced d-times before it becomes stationary. The method can be followed in formula (3), shown here for I(1) series.

$$y_t - y_{t-1} = y_{t-1} + u_t - y_{t-1} \Rightarrow \Delta y_t = u_t \quad (3)$$

Since  $u_t$  is stationary, so is the difference of  $y_t$ . “Stationary series are said to be integrated of order zero, I(0).” (Hill et al. 2008, p. 338).

Finally, one has to test whether the series are cointegrated. It implies that two processes, both of I(1) order, have similar stochastic trends and they never drift too far apart. Moreover, a linear difference of those processes is stationary. If the processes, thus, are  $y_t$  and  $x_t$ , while their difference is  $e_t$ , so that difference must be of order I(0). This aforementioned difference can be pictured with the help of the formula (4).

$$e_t = y_t - \beta_1 - \beta_2 \times x_t \quad (4)$$

Testing for cointegration is done by the means of testing for stationarity of the least squares residuals with Dickey-Fuller. The critical values here are different from the aforementioned

ADF test. That is because “(...) we are basing this test upon estimated values of the residuals (...)” (Hill et al. 2008, p. 339). It is also of big importance here to use the critical values in accordance to the used set of deterministic components. If the residuals are stationary, then the series are cointegrated. The null hypothesis, which stands here for no cointegration as well as the presence of unit root in the residuals, is rejected when the calculated  $\tau$  is smaller than the tabulated one.

All of the above steps and tests need to be done in order to create a proper time series model for data testing. The choice of the correct model is based on whether the cointegration occurs in or it is absent from the series. If the cointegration is present between the time series, then one ought to use the VEC (vector error correction) model. VAR (vector autoregressive) model is to be used in the situation when the null hypothesis of the lack of cointegration was not rejected. It has to be noted here that the “(...) VEC model is a special form of the VAR for I(1) variables that are cointegrated.” (Hill et al. 2008, p. 348). These two models in their equation forms are shown below in formulas (7) and (8):

$$\begin{aligned} \text{VAR model} \quad y_t &= \beta_{10} + \beta_{11} \times y_{t-1} + \beta_{12} \times x_{t-1} + v_t^y \\ x_t &= \beta_{20} + \beta_{21} \times y_{t-1} + \beta_{22} \times x_{t-1} + v_t^x \end{aligned} \quad (5)$$

$$\begin{aligned} \text{VEC model} \quad y_t &= \alpha_{10} + (\alpha_{11} + 1) \times y_{t-1} - \alpha_{11} \times \beta_0 - \alpha_{11} \times \beta_1 \times x_{t-1} + v_t^y \\ x_t &= \alpha_{20} + \alpha_{21} \times y_{t-1} - \alpha_{21} \times \beta_0 - (\alpha_{21} \times \beta_1 - 1) \times x_{t-1} + v_t^x \end{aligned} \quad (6)$$

It is quite clear that in the VAR model case the first variable is dependent on the past value of the other value as well as its own first lag. The second variable is, of course, a function of the lag of the other variable from the system and the first lag of itself. If the variables are non-stationary then they need to be exchanged with their corresponding first differences in the VAR framework.

In the VEC model, the  $\alpha$ -coefficients (with subscripts ‘ $_{11}$ ’ and ‘ $_{21}$ ’) are named as error correction coefficients as they illustrate the extent to which differences of the tested variables react to the cointegrating error ( $e_{t-1}$ ). For example, if  $\alpha_{21}$  is negative then  $\Delta x_t$  decreases. As the counter-effect, to correct the error, a positive  $\alpha_{11}$  guarantees that  $\Delta y_t$  rises. “Having the error correction coefficients less than one in absolute value ensures that the system is not explosive.” (Hill et al. 2008, p. 348).

### 3.3. Time series analysis – working with the data

Since working with and adjusting the data within time series analysis is somewhat harder and more time-consuming than it is when one performs cross-tabulation tests, so a separate section devoted to showing the reader the process of getting the results and models seems like a sound solution here.

The first step is to create a number of additional variables representing the cancer mortality ratios. This means that each cancer mortality variable for each gender in each country is divided by its respective general mortality counterpart. This choice is based on the fact that only such ratios capture the annual changes in both cancer mortality and general mortality. These ratios are also a way to register the fluctuations of yearly cancer deaths as a proportion of overall deaths per each year. In addition, it is only by the means of these ratios that both

cancer and all-cause mortality can be included in the models. Thus, twelve new variables are added to the basic dataset. It is of importance to note here that it is those ratios, in their stationary forms, that will henceforth function as dependent variables in this thesis<sup>4</sup>.

As a result, one can begin the time series testing. First, all of the variables need to be tested for unit root. This is done with the help of the Augmented Dickey Fuller test. This test is easily carried out in STATA; one just needs to consider the amount of necessary lags as well as the inclusion/exclusion of certain components.

In terms of the variables from the Polish part of the dataset, it is only the ratios for female and male lung cancer mortality that are non-stationary. In addition, the same can be said about the proxy variable for air pollution. All three turn out to be of order I(1). In terms of the Spanish data, the deterministic proxy variables for obesity and air pollution turn out to be of order I(1). Female and male lung cancer mortality ratios are unit root of order I(2) and in turn their sum (variable 'ltspr') is I(1). The summary is presented below in Figure 3.3.1.

**Figure 3.3.1.** Summary of the stationarity testing

Country	Variable	Lags	Test value	Critical value	Components
Poland	Lung male	0/3	-3,282/-5,928	-3,41/-3,41	T&C/T&C
	Lung female	0/1	5,500/-2,115	-1,94/-1,94	No/No
	Lung total	2	-4,232	-3,41	T&C
	Stomach male	1	-5,991	-1,94	No
	Stomach female	1	-5,223	-1,94	No
	Stomach total	3	-4,306	-1,94	No
	Smoking habits	0	-3,097	-2,86	C
	Obesity	0	-3,872	-2,86	C
	Air pollution	1/0	-1,493/-2,406	-1,94/-1,94	No/No
	Spain	Lung male	3/2/1	0,415/-2,941/-11,371	-1,94/-3,41/-1,94
Lung female		1/2/2	3,709/-1,246/-7,100	-1,94/-1,94/-1,94	No/No/No
Lung total		3/0	0,840/-7,669	-1,94/-2,86	No/Constant
Stomach male		0	-5,564	-3,41	T&C
Stomach female		1	-5,485	-1,94	No
Stomach total		0	-5,107	-3,41	T&C
Smoking habits		0	-6,009	-3,41	T&C
Obesity		0/0	2,641/-4,994	-1,94/-1,94	No/No
Air pollution		0/0	-2,736/-5,331	-3,41/-1,94	T&C/No

Note: In all columns - Symbol / is used to indicate the result differences between the non-differenced and differenced series. In the Component column - T&C stands for "trend + constant", C signifies "just constant", whereas No is used for "no components".

The level of significance is 5%.

In the next step tests for cointegration are performed. A set of additional interaction variables is created to aid in this part of the time series analysis procedure. These variables are created by multiplying each stationary cancer mortality ratio with the Chernobyl dummy variable. The purpose of these variables is to account for the impact of the Chernobyl fallout on human cancer mortality. Testing for cointegration is also carried out in STATA, but the whole procedure is a bit more difficult than the unit root testing. Firstly, the regression is done where

<sup>4</sup> For the list of all of the additionally created variables outside of the main dataset please refer to the Appendix (Table A.3.1).

each cancer mortality ratio is used a dependent variable. The list of the included independent variables for each respective country is as follows: necessary determining factor variables, the aforementioned interaction variable as well as the standalone Chernobyl dummy. This means that, e.g. for male lung cancer mortality in Poland the proper mortality ratio is used as the dependent variable, whereas smoking habits and air pollution are used as the determining factor variables. Moreover, the Chernobyl dummy variable is included along with the correct interaction term. On the other hand, when e.g Spanish female stomach mortality model is considered then smoking habits and obesity are used as the determining independent variables. In addition, the Chernobyl dummy variable and the proper interaction term are also included as the explanatory variables, while a proper mortality ratio is used as the dependent variable. These regressions are, as such, performed for twelve different models: two cancer types in two countries for two genders with the addition of their summed total. It has to be mentioned that all of the variables that are used here have to be in their stationary forms. The next step is to perform a Dickey-Fuller test on the residuals saved from each separate regression with the respect to the number of lags and included components.

The results of cointegration testing are quite straightforward – cointegration occurs when lung cancer mortality figures are used as independent variables. It holds true for both Poland and Spain as well as for both genders and their total sum. There is no cointegration for stomach cancer deaths variables, irrespective of the country and gender. The summary of the results is located in the table below (Figure 3.3.2).

**Figure 3.3.2.** Summary of the cointegration testing

Country	Variable	Test value	Critical value	Components	Cointegration
Poland	Lung male	-8,741	-2,76	No	Yes
	Lung female	-9,244	-2,76	No	Yes
	Lung total	-3,776	-2,76	No	Yes
	Stomach male	-2,712	-2,76	No	No
	Stomach female	-2,739	-2,76	No	No
	Stomach total	-2,687	-2,76	No	No
Spain	Lung male	-8,051	-2,76	No	Yes
	Lung female	-9,499	-2,76	No	Yes
	Lung total	-11,668	-2,76	No	Yes
	Stomach male	-2,455	-2,76	No	No
	Stomach female	-2,598	-2,76	No	No
	Stomach total	-2,438	-2,76	No	No

Note: In the Components column - T&C stands for “trend + constant”, C signifies “just constant”, whereas No is used for “no components”.

The level of significance is 5%.

It is, as such, quite clear that all of the lung cancer deaths series need to be tested with the VEC model as the cointegration exists there between them and their relevant independent variables. In turn, all of the stomach cancer mortality series will be included in the VAR model because of the lack of cointegration between them and their independent variables. This means that in each case six models need to be created, tested and interpreted.

### 3.4. VAR models

In this part of the paper Vector Autoregressive models for the part of the dataset that is not cointegrated will be estimated and tested. This means that the stomach cancer mortality models are examined in this part of the paper. In terms of estimation, the modeling process is quite easy as one only needs to find out the proper amount of lags that have to be included in the model. This is done with the help of Information Criteria along with a Likelihood Ratio test. Testing the models can be a bit harder as it focuses on model stability as well as normality in the residuals. This latter part is of big importance as the normality in the residuals implies a proper estimation of the model in question. “Recall that hypothesis tests and interval estimates for the coefficients rely on the assumption that the errors, and hence the dependent variable  $y$ , are normally distributed.” (Hill et al. 2008, p. 89). This is tested with the Jarque-Bera (J-B) test for normality in the residuals, which has a null hypothesis of normal distribution in the errors. It needs to be noted here that the necessary information about model estimation and its testing (for each single model) is presented in this section, whereas the models themselves are placed in the Results section.

All of the VAR models need four lags in their specifications in accordance to the Selection-Order Criteria. The tests for stability of these models return positive results as all of the eigenvalues for each model are located within the unit circle. As such, all six VAR models are stable. When the normal distribution of the standard residuals is considered, only the model representing Spanish female deaths appears to not fulfill this requirement. In all of the other models the residuals appear to be normally distributed as the values of the Jarque-Bera tests prevent one from rejecting the null hypothesis. For the outputs of the J-B, please refer to the figures placed in the Appendix (Figure A.3.2 – A.3.7). For the summary of all of the necessary test specifications and test results, please refer to the chart presented in Figure 3.4.1 below. It is not very surprising to see that all of the models require the same amount of lags and that all of them turn out to be stable and have their eigenvalues located in the unit circle, since all of these models are quite similar to one another.

**Figure 3.4.1.** Summary of VAR modeling specifications

<b>Country</b>	<b>Model</b>	<b>Lags</b>	<b>Model stability</b>	<b>Normal residuals</b>
Poland	Stomach male	4	Yes	Yes
	Stomach female	4	Yes	Yes
	Stomach total	4	Yes	Yes
Spain	Stomach male	4	Yes	Yes
	Stomach female	4	Yes	No
	Stomach total	4	Yes	Yes

### 3.5. VEC models

This section of the paper is dedicated to testing for VEC (Vector Error Correction) models. While all of the necessary information about model estimation and its testing (for each single model) is presented in this section, the models themselves can be found in the Results section. VEC-modeling requires the knowledge of the number of test ranks. These ranks can also be interpreted as the amount of underlying cointegration relationships that take place in the model. To calculate the number of these relationships one should use the Johansen test for cointegration. In addition, the number of lags is decided by using, as in the VAR cases, the Information Criteria test. VEC models, similarly as VAR models, need also to be tested for

residual normality; this is done, accordingly, with the help of the Jarque-Bera test for normality in the error terms.

For the summary of model specifications and test results, please refer to the chart placed in the Figure 3.5.1 below. In all of the models four lags are used; similarly as with the VAR models. On the other hand, the estimated models differ from one another in terms of the amount of test ranks (cointegration relationships). Polish male lung cancer mortality ratio along with the Spanish female one are tested for up to three relationships. In both countries total mortality is only tested with one test rank. In addition, all of the models appear to have normally distributed residuals as the results of the Jarque-Bera tests are too big to allow the rejection of the null hypothesis. For the outputs of the J-B test for each of the VEC models, please see the Appendix section (Figure A.3.8 – A.3.13).

**Figure 3.5.1.** Summary of VEC modeling specifications

<b>Country</b>	<b>Variable</b>	<b>Rank</b>	<b>Lags</b>	<b>Normal residuals</b>
Poland	Lung male	3	4	Yes
	Lung female	2	4	Yes
	Lung total	1	4	Yes
Spain	Lung male	2	4	Yes
	Lung female	3	4	Yes
	Lung total	1	4	Yes

## 4. Results

Why is mortality so important? First answer that one may stumble upon is that it is a part of a natural human life-cycle and that everybody sooner or later faces death. Mortality can also be defined as “(...) the process by which the members of the population are reduced by death (...)” (Hinde 1998, p. 2). Why should demographers be concerned with it? It is a part of the balancing equation:  $P_{t+1} = P_t + B_t - D_t + I_t - E_t$ . In this equation deaths are represented by  $D_t$ . In connection with births/fertility ( $B_t$ ), mortality creates a part of this equation that is referred to as the natural increase. This increase can be positive or negative depending on which of the two factors is greater than the other. In addition it is a well-known issue that mortality varies with “(...) age, sex, occupation, marital status and so on.” (Hinde 1998, p. 19). The author then adds that it is one of the main tasks within demography to try to measure and understand those differences.

As mentioned before, the research question of this paper is to see whether Chernobyl radiation fallout really does have a significantly big effect on human cancer mortality. The main focus is on examining cancer-related deaths in Poland, while Spain is used as a control group. Lung cancer is used as the one that is clearly caused by radiation, whereas its stomach counterpart is used for comparison purposes as its development is not caused by radionuclides. In addition, proxies for smoking habits, obesity and air pollution are used as the determining factors. The gathered data is examined in two ways: 1) age-specific cancer mortality figures are tested with Pearson chi-square methods; 2) general cancer mortality figures are examined with the help of times series methods and VAR and VEC models (deterministic factor variables are used in this part of the analysis).

### 4.1. Cross-tabulations – results

Pearson  $\chi^2$ -tests are performed with the help of statistical software called SPSS. In addition, checking for Pearson p-values and abnormally big residuals is being carried out. The null hypothesis here states that each variable has an equal amount of observations in both of the tested periods; this means that the reactor radiation is said to have no effect on cancer mortality in the whole “after”-period (if all other deterministic factors are held constant).

#### *Group 0-29 years*

The output for this group is presented in Figure 4.1.1 below. One can see the comparison between the countries in regard to the period, gender and, in turn, to the cancer type. In the case of every cancer as well as gender, the amount of deaths is smaller in the second examined time-period. The biggest value is in the first period of male lung cancer mortality (466). The biggest difference between the two periods is for male lung cancer mortality in Spain (544).

It becomes also quite clear that the residuals in the second periods are not exceeding the limit of two and therefore we cannot reject the null hypothesis of equal figures in both periods. This is confirmed by the p-values, which are equal to 0,878 for male lung cancer mortality and 0,244 for the stomach counterpart. For female cancer deaths the p-values are smaller: 0,064 for lung and 0,020 for stomach. It is worth mentioning that the biggest residuals in the second period are attained by women, with 1,4 for female stomach cancer mortality in Spain as the

biggest and 0,0 for lung cancer Spanish men. The correlation between lower p-values, smaller differences and higher standard residual values is quite clear here.

**Figure 4.1.1.** Contingency table for age-group 0-29 years

Cancer	Gender	Period	1966-1985	Count	Country		Total	
					Poland	Spain		
Lung	Men	1966-1985		Count	466	544	1010	
				Expected Count	467,4	542,6	1010,0	
				Std. Residual	,0	,1		
		1986-2005		Count	217	249	466	
				Expected Count	215,6	250,4	466,0	
				Std. Residual	,1	,0		
	Total		Count	683	793	1476		
			Expected Count	683,0	793,0	1476,0		
	Women	1966-1985		Count	214	249	463	
				Expected Count	224,8	238,2	463,0	
				Std. Residual	-,7	,7		
			1986-2005		Count	105	89	194
				Expected Count	94,2	99,8	194,0	
				Std. Residual	1,1	-1,1		
Total			Count	319	338	657		
			Expected Count	319,0	338,0	657,0		
Stomach		Men	1966-1985		Count	464	371	835
					Expected Count	454,5	380,5	835,0
					Std. Residual	,4	-,5	
			1986-2005		Count	181	169	350
				Expected Count	190,5	159,5	350,0	
				Std. Residual	-,7	,8		
	Total		Count	645	540	1185		
			Expected Count	645,0	540,0	1185,0		
	Women	1966-1985		Count	366	250	616	
				Expected Count	349,0	267,0	616,0	
				Std. Residual	,9	-1,0		
			1986-2005		Count	175	164	339
				Expected Count	192,0	147,0	339,0	
				Std. Residual	-1,2	1,4		
Total			Count	541	414	955		
			Expected Count	541,0	414,0	955,0		

### Group 30-59 years

The output for this group is presented in Figure 4.1.2 below. Once again one can see the comparison between the countries in regard to the period, gender and, last but not least, to the cancer type. The numbers are bigger than their respective counterparts in the younger age-group. In this group one can see that it is not always the case that the amount of deaths decreases between the periods. It actually holds true that mortality figures are greater in the case of the lung cancer. The biggest value (84 688) is for the second period of male lung cancer mortality. The biggest difference, in absolute terms, is for lung cancer among Spanish men (124713).

The residuals are clearly exceeding the limit of two for all lung cancer death cases. In the second period, the biggest residual value is for Spanish males who died of lung cancer (12,3). It is worth noting that for all stomach cancer women the residuals attain the value of zero. The p-values are as follows: men 0,000 for lung cancer and 0,053 for stomach one, women 0,000 for lung and 0,926 for stomach cancer. The correlation between the tested factors is yet again very clear. The null hypothesis of equal amount of observations in both time periods for all variables is rejected for lung cancer deaths among both genders. On the other hand, it cannot be rejected for stomach ones.

**Figure 4.1.2.** Cross-tabulation for age-group 30-59 years

Cancer	Gender	Period		Country		Total
				Poland	Spain	
Lung	Men	1966-1985	Count	61643	35270	96913
			Expected Count	58701,6	38211,4	96913,0
			Std. Residual	12,1	-15,0	
		1986-2005	Count	84688	59983	144671
			Expected Count	87629,4	57041,6	144671,0
			Std. Residual	-9,9	12,3	
	Total	Count	146331	95253	241584	
		Expected Count	146331,0	95253,0	241584,0	
	Women	1966-1985	Count	8999	5768	14767
			Expected Count	10025,0	4742,0	14767,0
			Std. Residual	-10,2	14,9	
		1986-2005	Count	19674	7795	27469
Expected Count			18648,0	8821,0	27469,0	
Std. Residual			7,5	-10,9		
Total	Count	28673	13563	42236		
	Expected Count	28673,0	13563,0	42236,0		
Stomach	Men	1966-1985	Count	32779	22503	55282
			Expected Count	32915,2	22366,8	55282,0
			Std. Residual	-,8	,9	
		1986-2005	Count	19631	13111	32742
			Expected Count	19494,8	13247,2	32742,0
			Std. Residual	1,0	-1,2	
	Total	Count	52410	35614	88024	
		Expected Count	52410,0	35614,0	88024,0	
	Women	1966-1985	Count	14400	11004	25404
			Expected Count	14404,3	10999,7	25404,0
			Std. Residual	,0	,0	
		1986-2005	Count	7731	5896	13627
Expected Count			7726,7	5900,3	13627,0	
Std. Residual			,0	,0		
Total	Count	22131	16900	39031		
	Expected Count	22131,0	16900,0	39031,0		

*Group 60-85+ years*

The output for this group is presented in Figure 4.1.3 below. One can, again, see the comparison between the countries in regard to the period, with gender and cancer type put as the layers. Numbers here are mostly greater than in the two aforementioned age-groups. Figures for the second examined period in this group are bigger again only for the lung cancer mortality. The biggest value is for the second period male lung cancer victims in Spain (216 518), while the biggest difference is also for the lung cancer mortality among men in Spain. In absolute terms its value reaches |11 7384|.

All of the residuals in this group are greater than two. The biggest post-Chernobyl residual value is attained by Spanish male lung cancer deaths (value of 14,5). When the smallest residual in the “after”-period is regarded, then it has to be noted that it occurs for stomach cancer victims among women in Spain (value of 4,9). The p-values are equal to 0,000 for all of the tested groups in this case. It seems only natural as the standard residuals do exceed the limit value of two, sometimes by a big difference. The relationship between lower p-values, smaller differences and higher standard residual values is again very visible. Since all of the p-values are equal to maximum three-star significance, then the null hypothesis of equal figures in both periods of time has to be rejected for all variables in this age-group.

**Figure 4.1.3.** Pearson chi-square table for age-group 60-85+ years

Cancer	Gender	Period		Country			
				Poland	Spain	Total	
Lung	Men	1966-1985	Count	114313	99134	213447	
			Expected Count	107671,5	105775,5	213447,0	
			Std. Residual	20,2	-20,4		
		1986-2005	Count	206997	216518	423515	
			Expected Count	213638,5	209876,5	423515,0	
			Std. Residual	-14,4	14,5		
	Total	Count	321310	315652	636962		
		Expected Count	321310,0	315652,0	636962,0		
	Women	Period	1966-1985	Count	21851	16289	38140
				Expected Count	23963,3	14176,7	38140,0
				Std. Residual	-13,6	17,7	
			1986-2005	Count	47165	24541	71706
Expected Count				45052,7	26653,3	71706,0	
Std. Residual				10,0	-12,9		
Total		Count	69016	40830	109846		
		Expected Count	69016,0	40830,0	109846,0		
Stomach		Men	1966-1985	Count	89037	77210	166247
				Expected Count	86469,6	79777,4	166247,0
				Std. Residual	8,7	-9,1	
			1986-2005	Count	60292	60562	120854
	Expected Count			62859,4	57994,6	120854,0	
	Std. Residual			-10,2	10,7		
	Total	Count	149329	137772	287101		
		Expected Count	149329,0	137772,0	287101,0		
	Women	Period	1966-1985	Count	57932	62115	120047
				Expected Count	56941,3	63105,7	120047,0
				Std. Residual	4,2	-3,9	
			1986-2005	Count	36073	42067	78140
Expected Count				37063,7	41076,3	78140,0	
Std. Residual				-5,1	4,9		
Total		Count	94005	104182	198187		
		Expected Count	94005,0	104182,0	198187,0		

### Total

The output for the sums of all age-groups is presented in the table that is located below (Figure 4.1.4.). This table represents the differences between the countries in regard to the period, gender and, in turn, to the cancer type. The numbers are bigger in the second period in comparison to the first one only when lung cancer is considered. They decrease in the period after the Chernobyl reactor meltdown in terms of stomach cancer. The biggest number is for post-Chernobyl Polish lung cancer deaths among men (value of 291 902). When one looks at the differences, the biggest difference, in absolute terms, is for male lung cancer mortality in Spain (141 802).

It is quite obvious that all of the residuals in this group should be greater than two. The biggest residual value in the second period is attained by the male lung cancer deaths in Spain (value of 20,6). It needs to be noted that the smallest value of the residuals occurs for stomach cancer victims among women in Spain (value of 5,3). The p-values are three-star significant for all of the variables in this case. The relationship that occurs between lower p-values, smaller differences and higher standard residual values is again quite clear. When the null hypothesis of equal figures in both periods of time is regarded, it has to be rejected for all variables in this age-group.

**Figure 4.1.4.** Cross-tabulation for the sums of the respective age-groups

Cancer	Gender	Period		Country			
				Poland	Spain	Total	
Lung	Men	1966-1985	Count	176422	134948	311370	
			Expected Count	165702,7	145667,3	311370,0	
			Std. Residual	26,3	-28,1		
		1986-2005	Count	291902	276750	568652	
			Expected Count	302621,3	266030,7	568652,0	
			Std. Residual	-19,5	20,8		
	Total	Count	468324	411698	880022		
		Expected Count	468324,0	411698,0	880022,0		
	Women	Period	1966-1985	Count	31064	22306	53370
				Expected Count	34245,9	19124,1	53370,0
				Std. Residual	-17,2	23,0	
			1986-2005	Count	66944	32425	99369
Expected Count				63762,1	35606,9	99369,0	
Std. Residual				12,6	-16,9		
Total		Count	98008	54731	152739		
		Expected Count	98008,0	54731,0	152739,0		
Stomach		Men	1966-1985	Count	122280	100084	222364
				Expected Count	119590,0	102774,0	222364,0
				Std. Residual	7,8	-8,4	
			1986-2005	Count	80104	73842	153946
	Expected Count			82794,0	71152,0	153946,0	
	Std. Residual			-9,3	10,1		
	Total	Count	202384	173926	376310		
		Expected Count	202384,0	173926,0	376310,0		
	Women	Period	1966-1985	Count	72698	73369	146067
				Expected Count	71555,8	74511,2	146067,0
				Std. Residual	4,3	-4,2	
			1986-2005	Count	43979	48127	92106
Expected Count				45121,2	46984,8	92106,0	
Std. Residual				-5,4	5,3		
Total		Count	116677	121496	238173		
		Expected Count	116677,0	121496,0	238173,0		

To sum up, the differences between the two time periods are significantly big for the age-group 60-85+ in both countries and for both types of cancers. For the group 30-59 the null hypothesis cannot be rejected for stomach cancer deaths in both Poland and Sweden, whereas the differences between two examined time periods seem to be statistically significant when lung cancer is regarded. In the youngest group (0-29) the null hypothesis of having equal amount of observations in each period cannot be rejected at all, regardless of the country, gender or the cancer type. In the case of the 'total', the calculated residuals are big enough to provide evidence against the null hypothesis. For the overview of the results of the Pearson chi-square cross-tabulations tests, please check the table below (Figure 4.1.5). It is quite clear, but needs to be nonetheless noted, that whenever the difference results in a negative value, it means that the amount of observations is greater in the second period (the difference is calculated in the following manner: "period before" – "period after").

**Figure 4.1.5.** Results of the cross-tabulation tests

Country	Cancer type	Gender	Age-group	Residuals	H <sub>0</sub> rejected?	Difference
Poland	Lung	Male	0-29	0,1	No	249
			30-59	-9,9	Yes	-23045
			60-85+	-14,4	Yes	-92684
			<i>Total</i>	-19,5	Yes	-115480
		Female	0-29	1,1	No	109
			30-59	7,5	Yes	-10675
			60-85+	10,0	Yes	-25314
			<i>Total</i>	12,6	Yes	-35880
	Stomach	Male	0-29	-0,7	No	283
			30-59	1,0	No	13148
			60-85+	-10,2	Yes	28745
			<i>Total</i>	-9,3	Yes	42179
		Female	0-29	-1,2	No	191
			30-59	0,0	No	6669
60-85+			-5,1	Yes	20859	
<i>Total</i>			-5,4	Yes	28719	
Spain	Lung	Male	0-29	0,0	No	295
			30-59	12,3	Yes	-24713
			60-85+	14,5	Yes	-117384
			<i>Total</i>	20,8	Yes	-141802
		Female	0-29	-1,1	No	160
			30-59	-10,9	Yes	-2027
			60-85+	-12,9	Yes	-8252
			<i>Total</i>	-16,9	Yes	-10119
	Stomach	Male	0-29	0,8	No	202
			30-59	-1,2	No	9392
			60-85+	10,7	Yes	16648
			<i>Total</i>	10,1	Yes	26242
		Female	0-29	1,4	No	86
			30-59	0,0	No	5108
60-85+			4,9	Yes	20048	
<i>Total</i>			5,3	Yes	25242	

Note: Null hypothesis is rejected when the residuals exceed the absolute value of 2

It is interesting to see that the null hypothesis is rejected more often in the cases of lung cancer mortality. What can be the reason of this? It is also interesting to see that the pattern of rejecting/accepting the null hypothesis is clearly the same in both countries - one country being affected by the Chernobyl radioactive fallout, while the other one is not. Are there any other deterministic factors at play besides radiation (e.g. lifestyle and environment) that can cause cancer mortality? In the next section, the data will be tested with the help of time series analysis and additional deterministic factors that can cause cancer to find the answers to these questions and doubts.

#### 4.2. Time series analysis – results

In this section the results of the time series models, Vector Autoregressive and Vector Error Correction ones, are presented and interpreted. In addition to standard outputs of the tested models, Granger causality is examined for VAR models and cointegrating equations are inspected for VEC models. Each cancer type in each country is presented firstly in terms of the total mortality and then in respect to each gender.

VAR models are presented and interpreted below:

##### *VAR model for total stomach cancer mortality in Poland*

It is quite clear from the output (Figure 4.2.1) that the tested variables do not have an impact on the cancer mortality in this model as most of the coefficients are not significant. However, mortality ratio is affected by its own fourth lag. The p-value in this case indicates three-star significance (0,000). In addition, the coefficient value is clearly positive. This means that the current mortality ratio increases by circa 0,84 if the past fourth unit value goes up by one.

**Figure 4.2.1.** VAR model for total stomach cancer mortality in Poland

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
stplr						
stplr L4.	.8427704	.0729823	11.55	0.000	.6997277	.9858131
stplrd L4.	-.2880964	.2056831	-1.40	0.161	-.6912279	.1150352
smokepl L4.	-4.85e-08	4.76e-08	-1.02	0.308	-1.42e-07	4.47e-08
fatpl L4.	-.0000506	.0000475	-1.07	0.286	-.0001437	.0000424
chernobyl L4.	.0060919	.004094	1.49	0.137	-.0019323	.014116
_cons	.0108015	.0056466	1.91	0.056	-.0002657	.0218687

##### Granger causality wald tests

Equation	Excluded	chi2	df	Prob > chi2
stplr	stplrd	1.9619	1	0.161
stplr	smokepl	1.0394	1	0.308
stplr	fatpl	1.1373	1	0.286
stplr	chernobyl	2.2141	1	0.137
stplr	ALL	12.359	4	0.015

In terms of Granger-causality, the results also turn out to be insignificant because their p-values are greater than the critical value for the 5%-level. This is not true when the variable 'All' is regarded as its p-value is equal to 0,015. Apparently, even though single variables do not Granger-cause changes in total stomach cancer mortality in Poland, a combination of them does that.

##### *VAR model for male stomach cancer mortality in Poland*

In the male stomach cancer case, the situation is quite similar to the one presented above. The fourth lag of the male mortality ratio for this cancer type is the only variable that appears to have an impact on the variable itself. The p-value is three-star statistically significant in this case and the coefficient has a positive direction. The interpretation is as follows: If the mortality ratio increases by one unit within the fourth lag, then it means an increase by around 0,89 for the current ratio value. In addition, the constant appears to be significant with p-value equal to 0,014. Output is shown in Figure 4.2.2 below.

**Figure 4.2.2.** VAR model for male stomach cancer mortality in Poland

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
smp1r						
smp1r L4.	.889619	.0679757	13.09	0.000	.756389	1.022849
smp1rd L4.	-.2427517	.1736471	-1.40	0.162	-.5830938	.0975904
smokep1 L4.	-8.25e-08	5.58e-08	-1.48	0.139	-1.92e-07	2.68e-08
fatp1 L4.	-.0000818	.0000576	-1.42	0.155	-.0001946	.000031
chernoby1 L4.	.0077183	.0048333	1.60	0.110	-.0017548	.0171914
_cons	.0160517	.006549	2.45	0.014	.003216	.0288874

Granger causality wald tests

Equation	Excluded	chi2	df	Prob > chi2
smp1r	smp1rd	1.9543	1	0.162
smp1r	smokep1	2.1876	1	0.139
smp1r	fatp1	2.0197	1	0.155
smp1r	chernoby1	2.5501	1	0.110
smp1r	ALL	20.027	4	0.000

The results of the Granger causality test are not significant, except for the case when all of the variables are combined into one (variable 'All'). The p-value is equal to 0,000. This means that male mortality is affected by a combination of all of the included variables and not by independently single variables as all other p-values are insignificant.

*VAR model for female stomach cancer mortality in Poland*

The results of the model representing stomach cancer deaths among Polish women closely follow the results of the previous two models. In the output (Figure 4.2.3), the coefficients are insignificant and, thus, their variables are not impacting the mortality ratio. This does not hold true, however, for the past value of the ratio itself. There is three-star significance in that case. This means that the present ratio goes up by almost 0,77 if the ratio four years back had increased in value by one unit.

**Figure 4.2.3.** VAR model for female stomach cancer mortality in Poland

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
sfp1r						
sfp1r L4.	.766943	.093451	8.21	0.000	.5837825	.9501035
sfp1rd L4.	-.3833829	.3001001	-1.28	0.201	-.9715683	.2048025
smokep1 L4.	-3.35e-08	4.69e-08	-0.71	0.475	-1.26e-07	5.85e-08
fatp1 L4.	-.0000201	.0000467	-0.43	0.667	-.0001116	.0000714
chernoby1 L4.	.00489	.0039129	1.25	0.211	-.0027792	.0125592
_cons	.0070279	.0056095	1.25	0.210	-.0039665	.0180223

Granger causality wald tests

Equation	Excluded	chi2	df	Prob > chi2
sfp1r	sfp1rd	1.632	1	0.201
sfp1r	smokep1	.50996	1	0.475
sfp1r	fatp1	.18476	1	0.667
sfp1r	chernoby1	1.5618	1	0.211
sfp1r	ALL	5.4924	4	0.240

All of the results are statistically insignificant, even for the variable ‘All’, when it comes to testing for Granger causality as all p-values in the output above are bigger than the critical value. As such, the cancer mortality ratio here is not Granger-caused by any of the variables included in this VAR model.

*VAR model for total stomach cancer mortality in Spain*

From the output presented below (Figure 4.2.4) it can be gathered that only the past fourth lag of the current cancer mortality ratio value is significant. It has a p-value equal to 0,000. In addition, the coefficient here has a positive direction. The interpretation of this coefficient is as follows: With an increase in value by one unit of the fourth lag of the mortality ratio itself, the current value increases by about 0,82. This means that the rest of the included variables do not affect the dependent variable.

**Figure 4.2.4.** VAR model for total stomach cancer mortality in Spain

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
stspr						
stspr L4.	.8166114	.0880013	9.28	0.000	.6441319	.9890908
stsprd L4.	-.1839749	.1811601	-1.02	0.310	-.5390423	.1710924
smokesp L4.	-1.88e-08	1.19e-08	-1.59	0.113	-4.21e-08	4.44e-09
dfatsp L4.	.0000434	.0000498	0.87	0.383	-.0000542	.000141
chernobyl L4.	.0028686	.0039848	0.72	0.472	-.0049416	.0106787
_cons	.0041293	.0028227	1.46	0.144	-.0014031	.0096617

Granger causality wald tests

Equation	Excluded	chi2	df	Prob > chi2
stspr	stsprd	1.0313	1	0.310
stspr	smokesp	2.5155	1	0.113
stspr	dfatsp	.75969	1	0.383
stspr	chernobyl	.51822	1	0.472
stspr	ALL	5.4163	4	0.247

In terms of Granger causality, all of the results appear to be non-significant as their p-values are bigger than the critical value for the 5%-level. Therefore, it can be stated that the Spanish total stomach cancer mortality ratio is not Granger-caused by any other variable included in this particular VAR model.

*VAR model for male stomach cancer mortality in Spain*

In the Spanish male case, the results are quite similar to those for the models above. For the output, please refer to Figure 4.2.5. The mortality ratio in this VAR is positively affected by the fourth its own lag value. There is three-star significance in this case. With a rise in value by one unit in the fourth lag ratio, the present ratio goes up by circa 0,85. Nevertheless, a significant coefficient also occurs when smoking habits are regarded (p-value of 0,036). The value of this coefficient is, however, negative and very low.

**Figure 4.2.5.** VAR model for male stomach cancer mortality in Spain

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
smspr						
smspr L4.	.8524019	.1103906	7.72	0.000	.6360402	1.068763
smsprd L4.	-.2200977	.2171339	-1.01	0.311	-.6456723	.2054768
smokesp L4.	-2.98e-08	1.42e-08	-2.10	0.036	-5.75e-08	-2.02e-09
dfatsp L4.	.0000462	.0000617	0.75	0.454	-.0000747	.000167
chernobyl L4.	.0049114	.0060498	0.81	0.417	-.0069459	.0167687
_cons	.0047936	.0041035	1.17	0.243	-.0032492	.0128363

Granger causality wald tests

Equation	Excluded	chi2	df	Prob > chi2
smspr	smsprd	1.0275	1	0.311
smspr	smokesp	4.4201	1	0.036
smspr	dfatsp	.55991	1	0.454
smspr	chernobyl	.65907	1	0.417
smspr	ALL	6.33	4	0.176

From Granger causality test, it can be also gathered that smoking habits are indeed impacting male cancer stomach mortality in Spain. The p-value here is equal, again, to 0,036. None of the other included variables seem to have an impact on the model's dependent variable as their p-values are insignificant.

*VAR model for female stomach cancer mortality in Spain*

The final VAR model also returns insignificant results (Figure 4.2.6). This is not true only for the past value of the dependent variable itself. The p-value is equal to 0,000 and according to the positive coefficient, an increase by one unit for the lag four ratio equal to a increase by circa 0,75 unit for the present ratio.

**Figure 4.2.6.** VAR model for female stomach cancer mortality in Spain

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
sfspr						
sfspr L4.	.75172	.0825999	9.10	0.000	.5898271	.9136128
sfsprd L4.	-.1037082	.1792392	-0.58	0.563	-.4550105	.2475942
smokesp L4.	-1.17e-08	1.17e-08	-1.00	0.317	-3.47e-08	1.12e-08
dfatsp L4.	.0000472	.0000475	1.00	0.320	-.0000458	.0001403
chernobyl L4.	.0004377	.0030109	0.15	0.884	-.0054636	.006339
_cons	.004105	.0022715	1.81	0.071	-.0003471	.0085571

Granger causality wald tests

Equation	Excluded	chi2	df	Prob > chi2
sfspr	sfsprd	.33478	1	0.563
sfspr	smokesp	1.0003	1	0.317
sfspr	dfatsp	.99023	1	0.320
sfspr	chernobyl	.02113	1	0.884
sfspr	ALL	6.0393	4	0.196

In addition, testing for causality in the model also brings insignificant results. None of the included variables appear to be Granger-causing the mortality ratio in this VAR model as their p-values are not significant.

In the next part of this section VEC models are presented and interpreted:

*VEC model for total lung cancer mortality in Poland*

From the output of this model (Figure 4.2.7) it can be seen that none of the three past lags of each of the included variables have a short-run impact on the mortality ratio as all of the respective p-values are insignificant. Only the constant in the VEC appears to be significant with a p-value of 0,036.

**Figure 4.2.7.** VEC model for total lung cancer mortality in Poland

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
d_ltp1r						
_ce1						
L1.	-.1217914	.1466161	-0.83	0.406	-.4091537	.165571
ltp1r						
LD.	-.4619894	.4800048	-0.96	0.336	-1.402782	.4788028
L2D.	-.0952707	.4282467	-0.22	0.824	-.9346187	.7440774
L3D.	.097487	.4453953	0.22	0.827	-.7754717	.9704458
ltp1rd						
LD.	-.6825652	1.094638	-0.62	0.533	-2.828016	1.462885
L2D.	.5031675	.959691	0.52	0.600	-1.377792	2.384127
L3D.	.7331543	.8608652	0.85	0.394	-.9541104	2.420419
smokepl						
LD.	-1.16e-07	1.42e-07	-0.82	0.414	-3.93e-07	1.62e-07
L2D.	-7.16e-08	1.11e-07	-0.64	0.520	-2.90e-07	1.47e-07
L3D.	-3.37e-08	8.10e-08	-0.42	0.677	-1.92e-07	1.25e-07
dso2pl						
LD.	.0005291	.0004645	1.14	0.255	-.0003814	.0014396
L2D.	.0004165	.0003516	1.18	0.236	-.0002727	.0011057
L3D.	.0002624	.0001923	1.36	0.172	-.0001146	.0006394
chernobyl						
LD.	.0332515	.0457694	0.73	0.468	-.0564547	.1229578
L2D.	-.013527	.034314	-0.39	0.693	-.0807812	.0537273
L3D.	-.0256919	.031905	-0.81	0.421	-.0882246	.0368408
_cons	.0016481	.0007858	2.10	0.036	.0001081	.0031882

beta	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
_ce1						
ltp1r	1	.	.	.	.	.
ltp1rd	-3.014721	.2722611	-11.07	0.000	-3.548343	-2.481099
smokepl	-9.92e-07	9.43e-08	-10.52	0.000	-1.18e-06	-8.07e-07
dso2pl	.0040752	.0004787	8.51	0.000	.003137	.0050134
chernobyl	.1628478	.0171834	9.48	0.000	.129169	.1965265
_cons	.0545222	.	.	.	.	.

The results for the long-run equilibrium (when the cointegrating equation is equal to zero) look a bit more promising. The constant in that case is not significant, but the rest of the variables return three-star significant results. The first case here is  $dltplr_t = 3,015 * ltp1rd_t$ . It can be, as such, constituted that there is a one unit rise in the interaction term which is equal to 3,015 unit increase of the mortality ratio itself. Next,  $dltplr_t = 9,92e^{-07} * smokepl_t$  - this means that as the amount of smoked cigarettes goes up by one unit, then the mortality ratio itself goes up by a small amount as well. In addition, from this equation  $dltplr_t = -0,004 * dso2pl_t$  it can be gathered that if the emissions of the SO<sub>2</sub> increase by one, then the mortality ratio decreases by about 0,0041.

In the case of the broken long-term equilibrium, which takes place when the cointegrating equation is different from zero, the error correction for the mortality ratio is negative and insignificant (for ce1 it is equal to about -0,12 with a p-value of 0,406). In turn, only the proxy for air pollution returns a significant result (coefficient: -816,7; p-value: 0,005). The coefficient for smoking habits is greatly positive and equal 499 754,7, but the p-value is insignificant (0,406). The coefficient z for the interaction terms is positive (0,81), but its p-

value is insignificant (0,472). It means that only the variable for SO<sub>2</sub> emissions is reacting to the cointegration error and adjusting if the equilibrium is broken. The other output mentioned in this paragraph is placed in the Appendix (Figure A.4.1).

#### VEC model for male lung cancer mortality in Poland

The situation in this VEC model is quite similar to the previous one. It is only the constant that turns out to be significant (p-value equal to 0,038), whereas none of the past values of any of the tested variables appear to have a short-run impact on the dependent variable. The output is presented in Figure 4.2.8 below.

**Figure 4.2.8.** VEC model for male lung cancer mortality in Poland

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
<b>d_d1mplr</b>						
_ce1						
L1.	-2.012947	1.271742	-1.58	0.113	-4.505515	.479621
_ce2						
L1.	-.2448959	3.088844	-0.08	0.937	-6.298918	5.809127
_ce3						
L1.	-1.87e-08	1.12e-07	-0.17	0.868	-2.39e-07	2.02e-07
<b>d1mplr</b>						
LD.	.5984492	1.243201	0.48	0.630	-1.83818	3.035078
L2D.	.8219199	1.052293	0.78	0.435	-1.240536	2.884376
L3D.	.6324757	.5122979	1.23	0.217	-.3716098	1.636561
<b>d1mplrd</b>						
LD.	-.1177626	2.339188	-0.05	0.960	-4.702487	4.466962
L2D.	-.4240046	1.717261	-0.25	0.805	-3.789775	2.941766
L3D.	-.0689201	.7463345	-0.09	0.926	-1.531709	1.393869
<b>smokep1</b>						
LD.	3.04e-08	1.76e-07	0.17	0.863	-3.14e-07	3.75e-07
L2D.	-7.82e-09	1.10e-07	-0.07	0.943	-2.24e-07	2.08e-07
L3D.	-5.23e-08	1.15e-07	-0.46	0.649	-2.77e-07	1.73e-07
<b>dso2p1</b>						
LD.	.0000759	.0003431	0.22	0.825	-.0005966	.0007484
L2D.	.000012	.0002312	0.05	0.959	-.0004412	.0004652
L3D.	.0000565	.0002443	0.23	0.817	-.0004222	.0005353
<b>chernoby1</b>						
LD.	.0025523	.003877	0.66	0.510	-.0050466	.0101512
L2D.	.0032057	.0032004	1.00	0.317	-.003067	.0094784
L3D.	.0005976	.0054432	0.11	0.913	-.0100709	.0112662
_cons	.0055071	.002656	2.07	0.038	.0003015	.0107128

beta	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
<b>_ce1</b>						
d1mplr	1	.	.	.	.	.
d1mplrd	4.44e-16	.	.	.	.	.
smokep1	(omitted)	.	.	.	.	.
dso2p1	-.0005777	.0002304	-2.51	0.012	-.0010292	-.0001262
chernoby1	-.0025901	.001442	-1.80	0.072	-.0054163	.0002361
_cons	-.0001663	.	.	.	.	.

The constant is insignificant in the cointegrating equations, while the proxy for air pollution turns out to be significant. The p-value here is equal to 0,012 and the proper equation look for when the equilibrium's cointegration is equal to zero is  $d1mplr_t = 0,0006 * dso2pl_t$ . This constitutes that a rise by about 0,0006 unit in the Polish male lung cancer deaths ratio is equal to the increase in air pollution by one unit. The p-value for the Chernobyl dummy variable is not significant as can be gathered from the above output.

When the cointegrating equations are not zero (broken equilibrium situation), then it can be seen that the error correction for the mortality ratio in the VEC model is insignificant (p-value of 0,113) with a negative coefficient. The error is not adjusted for by the air pollution variable as, even though its coefficient is positive (3376,8), the p-value is insignificant (0,155). The output mentioned here is located in Figure A.4.2 in the Appendix.

*VEC model for female lung cancer mortality in Poland*

In the case of this VEC model, the results may come as a bit of surprising as quite some coefficients appear to be significant in their short-term forms (output is located in Figure 2.4.9 below). The first coefficient is, of course, the constant with a p-value equal to 0,019. The second lag of the mortality ratio variable is also significant, but barely, with a p-value of 0,049 and a negative coefficient, which is equal to circa -2,55. This means that if the second lag increases by one unit, then the dependent variable in the present time decreases by -2,55 units. All three lags of the interaction term are significant. First lag has a p-value of 0,002 and a positive coefficient of 10,9; second lag's p-value is 0,005, while the positive coefficient is equal to 6,76; for the third lag the p-value is a bit higher and equal to 0,19 and the coefficient is lower, but still positive, and equal to 2,7. This constitutes that if the Chernobyl effect in each of its lags is increased by one unit, then the mortality ratio increases by about 10,9 and 6,76 as well 2,7, respectively. In addition, second lag of smoking habits is significant (0,47) and has a very small, but still positive, coefficient. This means that an increase in the number of smoked cigarettes has a small, albeit significant, impact of lung cancer mortality among women in Poland. All three lags of SO<sub>2</sub> are also greatly significant, with p-values that are equal to 0,001, 0,002 and 0,005 for each respective lag. The coefficients of all three lags are negative and their interpretation is as follows: If the emissions are increased by one in each of the past values, then this results in a mortality ratio decrease by 0,0005 for the first lag, 0,0004 for the second one and 0,0002 for the third lag.

**Figure 4.2.9.** VEC model for female lung cancer mortality in Poland

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
d_d1fp1r						
_ce1						
L1.	2.369195	2.045879	1.16	0.247	-1.640655	6.379045
_ce2						
L1.	-11.37235	3.640102	-3.12	0.002	-18.50682	-4.237883
d1fp1r						
LD.	-3.066207	1.738655	-1.76	0.078	-6.473908	.3414946
L2D.	-2.547646	1.296519	-1.96	0.049	-5.088777	-.0065146
L3D.	-1.170436	.724139	-1.62	0.106	-2.589722	.2488503
d1fp1rd						
LD.	10.88814	3.556777	3.06	0.002	3.916987	17.8593
L2D.	6.764908	2.424674	2.79	0.005	2.012634	11.51718
L3D.	2.732338	1.162379	2.35	0.019	.4541167	5.01056
smokep1						
LD.	4.04e-08	5.72e-08	0.71	0.479	-7.16e-08	1.53e-07
L2D.	8.96e-08	4.51e-08	1.98	0.047	1.11e-09	1.78e-07
L3D.	8.50e-08	5.27e-08	1.61	0.107	-1.84e-08	1.88e-07
dso2p1						
LD.	-.0005203	.0001554	-3.35	0.001	-.0008249	-.0002157
L2D.	-.0003699	.0001209	-3.06	0.002	-.0006068	-.000133
L3D.	-.0002317	.0000829	-2.80	0.005	-.0003941	-.0000693
chernobyl						
LD.	-.0094062	.0028692	-3.28	0.001	-.0150298	-.0037826
L2D.	-.0067414	.0022126	-3.05	0.002	-.0110779	-.0024048
L3D.	-.0052142	.001578	-3.30	0.001	-.008307	-.0021213
_cons	.0049729	.0021152	2.35	0.019	.0008272	.0091187

  

beta	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
_ce1						
d1fp1r	1	.	.	.	.	.
d1fp1rd	-1.11e-16	.	.	.	.	.
smokep1	1.49e-08	9.29e-09	1.60	0.110	-3.37e-09	3.31e-08
dso2p1	-.0000618	.0000286	-2.16	0.030	-.0001178	-5.83e-06
chernobyl	-.0006162	.0001768	-3.49	0.000	-.0009626	-.0002698
_cons	(omitted)					

By examining the cointegrating equations for long-term equilibrium (when equations are equal to zero) we see that air pollution, again, has a significant effect on the dependent variable. The equation is equal to  $dlfplr_t = 0,0006 * dso2pl_t$ , which is a similar result to the one in the male case. As such, the interpretation of the equation in that aforementioned model clearly applies here as well: A rise by about 0,0006 unit in the Polish female lung cancer deaths ratio is equal to the increase in air pollution by one unit. The constant is omitted in this model, while the p-value for smoking habits is insignificant.

When the equilibrium is broken and the equations are different from zero, then it is quite clear that the mortality ratio does not adjust for the error as the coefficient of about 2,4 is insignificant (p-value is equal to 0,247). Similar can be said about the air pollution variable, which has a negative coefficient (-4 377,216), but is also insignificant (0,392). For the other output, please see Figure A.4.3 in the Appendix.

#### VEC model for total lung cancer mortality in Spain

In the short-term equilibrium, all of the results of this VEC model are insignificant. This is even true for the constant. This means that there is no effect of the past values on the dependent variable. Output is located in the figure below (Figure 2.4.10).

**Figure 4.2.10.** VEC model for total lung cancer mortality in Spain

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
d_dltsp					
_ce1					
L1.	-1.83269	1.307766	-1.40	0.161	-4.395865 .7304845
dltsp					
LD.	.4129629	1.284074	0.32	0.748	-2.103776 2.929702
L2D.	.5288486	1.07322	0.49	0.622	-1.574624 2.632321
L3D.	.4906922	.5786286	0.85	0.396	-.643399 1.624783
dltspd					
LD.	-1.05301	1.381525	-0.76	0.446	-3.760748 1.654729
L2D.	-1.182896	1.255419	-0.94	0.346	-3.643472 1.277679
L3D.	-.5361388	.8133578	-0.66	0.510	-2.130291 1.058013
smokesp					
LD.	-7.39e-08	6.88e-08	-1.07	0.282	-2.09e-07 6.09e-08
L2D.	-9.55e-08	6.35e-08	-1.50	0.133	-2.20e-07 2.90e-08
L3D.	-1.41e-08	2.61e-08	-0.54	0.589	-6.53e-08 3.70e-08
dso2sp					
LD.	-.0003472	.0002453	-1.42	0.157	-.0008279 .0001335
L2D.	-.0002214	.0001772	-1.25	0.211	-.0005686 .0001259
L3D.	-.0000524	.0001113	-0.47	0.638	-.0002706 .0001658
chernoby1					
LD.	.0053593	.00556	0.96	0.335	-.0055381 .0162567
L2D.	.0008469	.006063	0.14	0.889	-.0110364 .0127303
L3D.	.0069693	.0061955	1.12	0.261	-.0051738 .0191123
_cons	.0001553	.0003077	0.50	0.614	-.0004478 .0007585

beta	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
_ce1					
dltsp					
dltspd	-.8930342	.068221	-13.09	0.000	-1.026745 -.7593235
smokesp	-5.29e-08	6.51e-09	-8.14	0.000	-6.57e-08 -4.02e-08
dso2sp	-.0002168	.0000344	-6.31	0.000	-.0002842 -.0001495
chernoby1	.0020499	.0001033	19.85	0.000	.0018475 .0022523
_cons	.0020588	.	.	.	.

In terms of the long-run equilibrium (when all equations are equal to zero), however, all of the included variables are significant. For the interaction variable the equation is  $dltsp_t = 0,893 * dltspd_t$ . This means that a when the value of the interaction terms goes up by one unit, then the mortality ratio itself increases by about 0,893. When smoking habits are regarded the equation is  $dltsp_t = 0,529e^{-08} * smokesp_t$ . According to this an increase by one

unit in the smoking habits leads to quite small, but still significant, increase in cancer mortality. Finally, when one looks at the air pollution emissions  $dlt spr_t = 0,0002 * dso2 sp_t$ ; this means that the mortality ratio increases by circa 0,0002 unit when the SO<sub>2</sub> emissions are increased by one unit. The constant, however, is not significant.

When the long-term error equilibrium is broken, then it can be seen that the mortality ratio is not correcting for the error as its p-value is insignificant (0,161). The coefficient here is equal to circa -1,83. However, the interaction ratio appears to be significant with a p-value of 0,034 and a coefficient -1,86. In addition, the proxy for smoking habits is also significant with a p-value of 0,05 and positive coefficient equal to  $3,49e^{07}$ . The proxy for air pollution returns insignificant results (p-value equals to 0,845), even though the coefficient here is greatly positive and equal to about 839,9. This means that, even though the mortality ratio in this model is not adjusting for the error, but other variables included in that particular model (except for the air pollution proxy variable) are doing that correction. Please see Figure A.4.4 in the Appendix section for the additional output.

### VEC model for male lung cancer mortality in Spain

From the output below (Figure 4.2.11) it can be seen that while most of the variables are insignificant in the short-run equilibrium, the significant results are for all three lags of the air pollution proxy. The constant is, yet again, also significant (0,000). In terms of all three lags of the 'dso2sp' variable, the coefficients are -0,0014, -0,00096 and -0,00052, respectively. The p-values are, in turn, equal to 0,001, 0,002 and 0,015, for the respective lags. This means that if the SO<sub>2</sub> emission increase by one unit, the mortality ratio in this VEC model goes down by 0,0014 in the first lag, by 0,00096 in the second and by 0,00052 in the third lag.

**Figure 4.2.11.** VEC model for male lung cancer mortality in Spain

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
d_d21mspr						
_ce1						
L1.	-5.228882	1.973477	-2.65	0.008	-9.096827	-1.360937
_ce2						
L1.	-.3505407	1.998403	-0.18	0.861	-4.267339	3.566257
d21mspr						
LD.	2.484859	1.70679	1.46	0.145	-.8603878	5.830106
L2D.	.8754615	1.106302	0.79	0.429	-1.29285	3.043773
L3D.	.1594942	.4427809	0.36	0.719	-.7083405	1.027329
d21msprd						
LD.	.6661518	1.750855	0.38	0.704	-2.765462	4.097765
L2D.	.7909947	1.194954	0.66	0.508	-1.551072	3.133062
L3D.	.6401386	.5566815	1.15	0.250	-.4509372	1.731214
smokesp						
LD.	4.93e-09	9.76e-08	0.05	0.960	-1.86e-07	1.96e-07
L2D.	-1.29e-07	8.18e-08	-1.57	0.116	-2.89e-07	3.16e-08
L3D.	8.14e-09	5.02e-08	0.16	0.871	-9.02e-08	1.06e-07
dso2sp						
LD.	-.001373	.0004166	-3.30	0.001	-.0021894	-.0005566
L2D.	-.0009611	.0003042	-3.16	0.002	-.0015573	-.000365
L3D.	-.0005246	.0002155	-2.43	0.015	-.000947	-.0001021
chernoby1						
LD.	.0164269	.0085159	1.93	0.054	-.0002639	.0331178
L2D.	-.0003932	.0116098	-0.03	0.973	-.0231479	.0223616
L3D.	.0207606	.0094916	2.19	0.029	.0021574	.0393638
_cons	.0216232	.0051455	4.20	0.000	.0115382	.0317083

beta	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
_ce1						
d21mspr	1	.	.	.	.	.
d21msprd	(omitted)	.	.	.	.	.
smokesp	-2.51e-08	2.13e-08	-1.18	0.238	-6.69e-08	1.66e-08
dso2sp	-.0003518	.0001017	-3.46	0.001	-.0005511	-.0001524
chernoby1	.0005238	.0002853	1.84	0.066	-.0000354	.001083
_cons	.0053243	.	.	.	.	.

In the long-run equilibrium (when all of the cointegrating equations are equal to zero), the significant results is delivered by the air pollution variable. The p-value is 0,001, while the coefficient is equal to circa -0,00035. The equation, as such, is  $d21mspr_t = 0,00035 * dso2sp_t$ . This constitutes that an increase by one unit in the sulfur dioxide emissions increases the Spanish male lung cancer mortality ratio by 0,00035 unit. The constant is not significant and so are the other p-values that are mentioned in the above output.

When the equation is not equal to zero as it is in the broken equilibrium case, the mortality ratio is surely significant (0,008) and has a negative coefficient of -5,23. However, the error is not picked up and corrected by the air pollution variable as, even though its coefficient is positive (about 3493), but its p-value is insignificant and equal to 0,277. This means that the mortality ratio variable acts up as the adjusting indicator in this model, whereas the SO<sub>2</sub> emissions one does not. For the other output connected to this model, please see the Figure A.4.5 in the Appendix.

#### VEC model for female lung cancer mortality in Spain

The final VEC model is for Spanish female mortality with lung cancer as its cause. In the short-run equilibrium none of the past lags of the included variables turn out to be significant; it is even true for the constant. This means that the independent variables included in the model have no effect on the dependent variable. The output is located in the table below (Figure 4.2.12).

**Figure 4.2.12.** VEC model for female lung cancer mortality in Spain

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
d_d21fspr						
_ce1						
L1.	-3.370944	3.384387	-1.00	0.319	-10.00422	3.262332
_ce2						
L1.	-1.865661	3.704286	-0.50	0.615	-9.125928	5.394607
_ce3						
L1.	5.10e-09	2.58e-08	0.20	0.843	-4.55e-08	5.57e-08
d21fspr						
LD.	.869349	2.83162	0.31	0.759	-4.680523	6.419221
L2D.	-.1494463	1.679147	-0.09	0.929	-3.440515	3.141622
L3D.	-.1230462	.5394288	-0.23	0.820	-1.180307	.9342148
d21fsprd						
LD.	2.157624	3.162026	0.68	0.495	-4.039834	8.355081
L2D.	1.734374	2.068389	0.84	0.402	-2.319595	5.788342
L3D.	.4195596	.9437557	0.44	0.657	-1.430167	2.269287
smokesp						
LD.	-2.93e-08	2.48e-08	-1.18	0.238	-7.78e-08	1.93e-08
L2D.	-1.58e-09	2.63e-08	-0.06	0.952	-5.32e-08	5.00e-08
L3D.	-4.49e-09	9.75e-09	-0.46	0.645	-2.36e-08	1.46e-08
dso2sp						
LD.	-.0000903	.0001491	-0.61	0.545	-.0003826	.0002019
L2D.	-.0000552	.0001243	-0.44	0.657	-.0002989	.0001885
L3D.	.0000161	.0000863	0.19	0.852	-.0001531	.0001852
chernoby1						
LD.	-.0007647	.0022498	-0.34	0.734	-.0051742	.0036449
L2D.	-.0009686	.002001	0.48	0.628	-.0029534	.0048905
L3D.	-.0009502	.0024764	-0.38	0.701	-.0058038	.0039035
_cons	-.0002874	.0002263	-1.27	0.204	-.0007309	.0001561

beta	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
_ce1						
d21fspr	1	.	.	.	.	.
d21fsprd	(omitted)					
smokesp	(omitted)					
dso2sp	-.0000106	.0000186	-0.57	0.567	-.0000471	.0000258
chernoby1	-.000056	.0000426	-1.31	0.189	-.0001395	.0000275
_cons	-.0000561	.	.	.	.	.

Similar can be said about the long-run equilibrium. None of the variables appear to have an effect on the mortality ratio in question as all of the p-values that are visible in the output are insignificant. This may be caused by some problems with the model and with the variable structure; this is confirmed by the fact that only this model lacks normally distributed residuals.

## 5. Discussion

The accident that took place on the 26th April, 1986, in Chernobyl, Ukraine, has changed the world and its view on nuclear energy. The consequences of the core meltdown in the reactor number four are still estimated as the biggest and most dangerous for human health, especially in terms of the delayed effects of the radiation fallout. Up to the recent times, the accident in the Nuclear Power Plant in Chernobyl was assessed as the most threatening. Only the current accident in Japan in the nuclear power plant located in Fukushima was evaluated to be of the same danger level to humanity as Chernobyl. Does this mean that all of the nuclear power plants should be shut down? The answer to that question is not easy and definitely not straightforward. As mentioned before, there are two sides to the nuclear energy argument. By asking both sides for help in answering that question, one will surely receive very mixed messages. However, at the present Poland is creating projects for its own nuclear power plant in the northern part of the country<sup>5</sup>. It is estimated to be completed by the year 2020.

Radiation causes many health problems for people who are exposed to it. Sometimes the effects on health are immediate, whereas at times they can take up to several years to develop. This is based on the doses as well as types of radiation. Apart from various diseases and mutations because of the exposure, human beings are also apt to be the victims of different types of cancer. The most prevalent types that are connected to contact with radionuclides are thyroid and leukemia, but a plethora of other types follows. Among these types lung cancer is often mentioned in the literature on radiation. The aim of this thesis, as mentioned before, is to test whether the Chernobyl radiation really has an effect on the development of lung cancer among people.

From the cross-tabulation part of the analysis, it can be gathered that there is no difference between the examined countries when it comes to rejecting the null hypothesis. More elderly die of those two cancer types in the period after the catastrophe. The middle age-group also seems to be more prone to die but only when the lung cancer is considered. On the whole, it is clear from these tests that lung cancer mortality is greater in both countries than the stomach cancer one as there are more negative results in the lung cancer part of “Difference” column of the summarizing chart (as shown in Figure 4.1.5). However, according to the calculated differences between the first and the second time period, it is Spain that has bigger mortality because of both cancer types. The results can be striking as a bit uncanny in the light of the fact that Spain was not within the reach of the Chernobyl’s radioactive cloud. In addition, it is also true that there are other factors that can have a delayed effect on health and mortality, besides smaller and non-acute doses of radiation. Chain-smoking, living in the areas with heavily polluted air or unhealthy eating do not have immediate effects on health or mortality. They may take years to actually impact human health and well-being; cancer on its own is also known to take time to develop. One may start smoking by the age of 18 only to be diagnosed with lung cancer when he or she turns 60. That is why these deterministic factors are used in the time series testing.

All of the VAR models, which are in fact all models with stomach cancer mortality as the dependent variable, support the hypothesis that Chernobyl is not the cause of increased mortality in the time period after the accident. The fourth lag of the interaction term in all

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<sup>5</sup> This fact is given here courtesy of my parents.

models is insignificant. In addition, all other lagged variables, except the past value of the dependent variables themselves, return insignificant results as well. This means that the hypothesis that stomach cancer is caused by excessive smoking or excessive eating is not supported here. When one, on the other hand, looks at Granger causality it becomes clear that in the male model for Poland the mortality ratio is impacted (Granger-caused) by a combination of all of the included variables. The reason for this might be that since the researched time period is quite short, then the effect of single factors on mortality cannot be registered by the time series analysis. What is registered, however, is the combined effect of the included variables. Similar can be said about the total model in Poland. Smoking habits appear to Granger-cause cancer mortality among men in Spain, whereas other variables do not. People in Southern Europe are known for their heavy smoking habits and men, in general and not only that in particular region, tend to have greater nicotine intakes than women. As such, they may be more apt to be diagnosed with cancer.

VEC models are created for lung cancer mortality data. In the short-run equilibrium only the Polish female model and Spanish male model return a number of significant results. Polish female mortality appears to be negatively affected by the Chernobyl radiation and by, to a smaller degree, smoking habits. The SO<sub>2</sub> emissions appear to be, in fact, decreasing the mortality ratio for Polish females. The effect of the latter factor is similar when Spanish males are considered – the rise in the amount of the emissions seems to be helping with the decrease of the mortality ratio. In the long-run equilibrium, however, the effect of air pollution clearly becomes negative in these two models. As such, the increasing amount of air pollutants results in increased lung cancer mortality. Moreover, the same can be said about the models representing Polish males and Spanish total. The reason for this is, as mentioned before, that there exists a delayed effect on health and that only the long-run analysis registers this as a negative impact. This means that both genders in Poland along with Spanish men are prone to be affected by air pollution. The explanation for Poland is quite simple – as the country became independent, the number of purchased and driven cars as well as the number of big factories increased rapidly over a short period of time. This obviously resulted in an increase in air pollution. Chernobyl has a negative long-run effect on mortality in total models for both countries, but the effect is much greater in Poland; 3,015 as opposed to Spanish 0,893. The reason for this result is that even though Spain was not affected by the radioactive cloud, the effects of the contamination might have also reached out into this country, via e.g. travelers, exported goods, etc. Poland's effect is bigger as this country is the neighboring nation of Ukraine. The reason for only the totals being impacted is that in those two models there are the biggest amounts of annual observations. It is, hence, easier for the time series analysis to register the causality factor. Similar can be said about the smoking habits. The effect of this habit on mortality is also only picked up in the total models for both Spain and Poland. The effect value for Poland is bigger than then one for Spain, although both of them are causing an increase in the respective mortality ratios, for Poland it is  $9,92e^{-07}$  as opposed to the Spanish  $0,529e^{-08}$ . These numbers in their decimal forms are roughly equal to 0,009 and 0,0002, respectively. This means that in the long-run Polish people are more affected by heavy smoking and tend to die of lung cancer more often than Spanish people.

There are also certain issues with this paper that need to be discussed. The development of the lung cancer is, of course, much less common among people who were exposed to the fallout in comparison to the thyroid one. That is why the literature on the health effects of radiation usually deals with the thyroid cancer - more data, easier detection of causality and better results are just a few reasons. The fact that lung cancer is less common among radiation

victims might also be the explanation for the fact that results of the Chernobyl radiation effect are only conclusive and significant in the total models in this paper.

On the other hand, some problems with the scarcity of the significant results may be due to the fact that all twelve models researched in this thesis are quite simple. In addition, the choice and the structure of the included variables might also be impacting the results. It is especially pertinent when the deterministic variables are regarded. None of these variables is direct as all of them are used as proxies. One may then argue about whether or not they are good proxies and even their relevance to the research may be questioned. The results may also be impacted by a clear lack of the variable that describes the yearly radiation doses. The latter issue cannot be resolved, however, due to the lack of open availability of such data.

It is furthermore true that in time series analysis one should use much longer periods than 40 years. The more annual observations are used, the better and easier the causality can be picked up and registered by the models. This is especially important when dealing with non-stationary variables that need to be differenced. Taking differences decreases the amount of yearly observations in the variables. This is especially an apparent issue for all of the deterministic factor variables as none of them covered the whole researched time period to begin with. Nevertheless, there were also hindrances from extending the dataset for both time periods. Mortality by each of the two cancers as the cause covers the period 1951-2005 for Spain and 1959-2005 for Poland. General mortality, in turn, spans the period 1908-2006 for Spain and 1958-2009 for Poland. I believe that I have managed to find a middle-ground for the examined time periods as well as for making these periods relevant to the topic of this paper and to my research question.

## 6. Conclusions

The accident in the Chernobyl's Nuclear Power Plant altered our view on the nuclear power. It needs to be added that the hazards that this accident brought into the world have actually changed us. Exposure to radionuclides can be harmful to human health and can result in immediate or delayed diseases, malfunctions or mutations. One of the most common consequences of being exposed to nuclear fallout is the development of cancer. A number of books and articles state that it is the thyroid and leukemia cancers that are the most common results of exposure to radioactivity. Many authors analyze just these two cancer types when writing about the consequences of the Chernobyl's reactor meltdown. Nonetheless, there are also other types of cancers that are caused by radionuclides and lung cancer is one them. In this paper the main aim is to test the hypothesis whether the Chernobyl radiation is still causing the development of lung cancer. The main part of the analysis uses the time series methods to test the significance of that hypothesis by creating VEC and VAR models.

To sum up the VEC model results, air pollution affects the mortality of Polish and Spanish males as well as Polish females in terms of lung cancer deaths. In addition, it is both the Chernobyl effect, which is conveyed by the interaction terms, as well as the heavy smoking habits that are resulting in an increased mortality in the total models in both countries. To sum up the VAR model results, total and male stomach cancer mortality ratios in Poland are impacted by a combination of all variables. In the case of the Spanish stomach cancer male mortality ratio it is mainly the excessive smoking that is causing more cancer deaths.

The results show that the radioactive contamination coming from the Chernobyl's accident is still affecting lung cancer mortality of people in Poland. However, it is the polluted air and even heavy smoking that pose much larger threats to human health nowadays. One cannot deny the influence of the radioactive fallout on the frequency of lung cancer diagnosis among Polish citizens, but it seems that it is the living in our modern and stressful times that is really taking its toll on lung cancer mortality. For the future research on this topic, one can only wish for longer time periods, different cancer types. Moreover, future papers on this or similar topic can really make use of more reliable and differently structured data. It should also be mentioned, that researchers should gain easier access to data on annual national radiation doses, which would really help and simplify their data analysis and modeling.

## 7. Sources

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## Data sources

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(*stomach/lung cancer mortality in Poland and Spain + age-specific cancer mortality for both countries and cancer types for groups 0-29/30-59/60-85+*) (accessed: 08.04.2011)
3. Smoking habits - <http://data.euro.who.int/hfad/>  
(*consumed cigarettes in million pieces per year*) (accessed: 11.04.2011)
4. Fat intake – <http://data.euro.who.int/hfad/>  
(*available fat per person per day in grams*) (accessed: 08.05.2011)
5. SO<sub>2</sub> emissions – <http://data.euro.who.int/hfad/>  
(*sulfur dioxide emissions per capita per year in kilograms*) (accessed: 08.05.2011)

## 8. Appendix

**Table A.2.1.** List of used variables and their explanations – for POLAND

<b>Variable name</b>	<b>Description</b>
year	Time period (1966-2005)
mpl, fpl, tpl	General mortality for males, females and resp. total in Poland
lmpl, lfpl, ltpl	Lung cancer mortality for males, females and resp. total in Poland
smpl, sfpl, stpl	Stomach cancer mortality for males, females and resp. total in Poland
smokepl	Millions of consumed cigarettes per year in Poland (proxy of smoking habits)
fatpl	Available fat per person per day in grams in Poland (proxy of obesity)
so2pl	SO <sub>2</sub> emission in Poland per capita per year in kilograms (proxy of air pollution)
ch	Dummy variable for “before” and “after” periods

**Table A.2.2.** List of used variables and their explanations – for SPAIN

<b>Variable name</b>	<b>Description</b>
year	Time period (1966-2005)
msh, fsh, tsh	General mortality for males, females and resp. total in Spain
lmsp, lfsp, ltsp	Lung cancer mortality for males, females and resp. total in Spain
smsp, sfsp, stsp	Stomach cancer mortality for males, females and resp. total in Spain
smokesp	Millions of consumed cigarettes per year in Spain (proxy of smoking habits)
fatsp	Available fat per person per day in grams in Spain (proxy of obesity)
so2sp	SO <sub>2</sub> emission in Spain per capita per year in kilograms (proxy of air pollution)
ch	Dummy variable for “before” and “after” periods

**Table A.3.1.** List of additionally created variables in the time series analysis section

Variable name	Description
lmspr, lfspr, ltspr	Ratios of lung cancer mortality in Spain for men, women and resp. total.
lmplr, lfplr, ltplr	Ratios of lung cancer mortality in Poland for men, women and resp. total.
smspr, sfspr, stspr	Ratios of stomach cancer mortality in Spain for men, women and resp. total.
smplr, sfplr, stplr	Ratios of stomach cancer mortality in Poland for men, women and resp. total.
dlnplr, dlfpplr	First difference of the male and resp. female lung cancer mortality ratios in Poland
dso2pl, dso2sp	First difference of the Polish and resp. Spanish air pollution proxies
d2lmspr, d2lfspr	Second difference of the male and resp. female stomach cancer mortality ratios in Spain
dfatp	First difference of the Spanish obesity proxy
dlnplr, dlfpplr, ltplr, smplr, sfplr, stplr, d2lmspr, d2lfspr, dltspr, smspr, sfspr, stspr	Interaction terms between each of the stationary cancer mortality ratios and the Chernobyl dummy

**Figure A.3.2.** Jarque-Bera test in VAR modeling for male stomach cancer mortality in Poland

Jarque-Bera test

Equation	chi2	df	Prob > chi2
smplr	2.620	2	0.26986
smplr	0.950	2	0.62197
smspr	1.926	2	0.38181
fatp	5.661	2	0.05897
chernobyl	1.601	2	0.44916
ALL	12.757	10	0.23756

Skewness test

Equation	Skewness	chi2	df	Prob > chi2
smplr	-.75952	2.596	1	0.10714
smplr	-.03048	0.004	1	0.94844
smspr	-.3916	0.690	1	0.40614
fatp	-.84102	3.183	1	0.07441
chernobyl	-.31471	0.446	1	0.50439
ALL		6.919	5	0.22675

Kurtosis test

Equation	kurtosis	chi2	df	Prob > chi2
smplr	3.1454	0.024	1	0.87740
smplr	3.9168	0.946	1	0.33086
smspr	4.048	1.236	1	0.26632
fatp	4.4843	2.478	1	0.11542
chernobyl	4.0133	1.155	1	0.28249
ALL		5.838	5	0.32226

**Figure A.3.3.** Jarque-Bera test in VAR modeling for female stomach cancer mortality in Poland

Jarque-Bera test

Equation	chi2	df	Prob > chi2
sfplr	0.141	2	0.93200
sfplrd	8.796	2	0.01230
smokepl	5.836	2	0.05405
fatpl	0.919	2	0.63153
chernoby1	3.309	2	0.19121
ALL	19.001	10	0.04026

Skewness test

Equation	Skewness	chi2	df	Prob > chi2
sfplr	.08621	0.033	1	0.85490
sfplrd	-.1339	0.081	1	0.77638
smokepl	-.77048	2.671	1	0.10217
fatpl	-.16503	0.123	1	0.72628
chernoby1	.78896	2.801	1	0.09420
ALL	5.709	5	0.33556	

Kurtosis test

Equation	Kurtosis	chi2	df	Prob > chi2
sfplr	3.309	0.107	1	0.74313
sfplrd	5.7833	8.715	1	0.00316
smokepl	4.6771	3.164	1	0.07526
fatpl	3.8415	0.797	1	0.37209
chernoby1	3.6718	0.508	1	0.47614
ALL	13.291	5	0.02080	

**Figure A.3.4.** Jarque-Bera test in VAR modeling for total stomach cancer mortality in Poland

Jarque-Bera test

Equation	chi2	df	Prob > chi2
stplr	0.364	2	0.83352
stplrd	8.526	2	0.01339
smokepl	2.880	2	0.23692
fatpl	2.138	2	0.34330
chernoby1	0.325	2	0.84982
ALL	14.334	10	0.15829

Skewness test

Equation	Skewness	chi2	df	Prob > chi2
stplr	-.28234	0.359	1	0.54921
stplrd	-.0803	0.029	1	0.86474
smokepl	-.51332	1.186	1	0.27619
fatpl	-.57008	1.462	1	0.22654
chernoby1	-.08712	0.034	1	0.85337
ALL	3.070	5	0.68917	

Kurtosis test

Equation	Kurtosis	chi2	df	Prob > chi2
stplr	2.9303	0.005	1	0.94111
stplrd	5.7644	8.597	1	0.00337
smokepl	4.2272	1.694	1	0.19304
fatpl	3.7751	0.676	1	0.41103
chernoby1	3.5089	0.291	1	0.58938
ALL	11.264	5	0.04639	

**Figure A.3.5.** Jarque-Bera test in VAR modeling for male stomach cancer mortality in Spain

Jarque-Bera test

Equation	chi2	df	Prob > chi2
smspr	0.788	2	0.67430
smsprd	8.263	2	0.01590
smokesp	12.972	2	0.00152
dfatsp	0.388	2	0.82363
chernoby1	0.679	2	0.71222
ALL	23.110	10	0.01035

Skewness test

Equation	Skewness	chi2	df	Prob > chi2
smspr	-.02209	0.002	1	0.96332
smsprd	-.46065	0.920	1	0.33760
smokesp	-.97223	4.096	1	0.04299
dfatsp	-.29849	0.386	1	0.53436
chernoby1	.27828	0.336	1	0.56239
ALL	5.739	5	0.33242	

Kurtosis test

Equation	Kurtosis	chi2	df	Prob > chi2
smspr	2.1482	0.786	1	0.37530
smsprd	5.6071	7.363	1	0.00666
smokesp	5.8624	8.876	1	0.00289
dfatsp	2.9573	0.002	1	0.96457
chernoby1	2.4372	0.343	1	0.55802
ALL	17.371	5	0.00385	

**Figure A.3.6.** Jarque-Bera test in VAR modeling for female stomach cancer mortality in Spain

Jarque-Bera test

Equation	chi2	df	Prob > chi2
sf spr	52.330	2	0.00000
sf sprd	0.733	2	0.69321
smokesp	22.403	2	0.00001
dfat sp	0.057	2	0.97174
chernobyl	2.893	2	0.23545
ALL	78.416	10	0.00000

Skewness test

Equation	Skewness	chi2	df	Prob > chi2
sf spr	2.0143	17.583	1	0.00003
sf sprd	-.10642	0.049	1	0.82467
smokesp	-1.3412	7.795	1	0.00524
dfat sp	.07849	0.027	1	0.87022
chernobyl	.72802	2.297	1	0.12965
ALL		27.750	5	0.00004

Kurtosis test

Equation	Kurtosis	chi2	df	Prob > chi2
sf spr	8.6635	34.748	1	0.00000
sf sprd	3.7945	0.684	1	0.40829
smokesp	6.6722	14.609	1	0.00013
dfat sp	2.8318	0.031	1	0.86104
chernobyl	3.7416	0.596	1	0.44019
ALL		50.666	5	0.00000

**Figure A.3.7.** Jarque-Bera test in VAR modeling for total stomach cancer mortality in Spain

Jarque-Bera test

Equation	chi2	df	Prob > chi2
st spr	3.718	2	0.15580
st sprd	3.836	2	0.14688
smokesp	15.981	2	0.00034
dfat sp	0.178	2	0.91486
chernobyl	2.188	2	0.33481
ALL	25.902	10	0.00387

Skewness test

Equation	Skewness	chi2	df	Prob > chi2
st spr	.78898	2.697	1	0.10051
st sprd	-.24627	0.263	1	0.60819
smokesp	-1.1842	6.077	1	0.01369
dfat sp	-.1609	0.112	1	0.73767
chernobyl	.70355	2.145	1	0.14304
ALL		11.295	5	0.04584

Kurtosis test

Equation	Kurtosis	chi2	df	Prob > chi2
st spr	3.9708	1.021	1	0.31230
st sprd	4.8162	3.574	1	0.05871
smokesp	6.0235	9.904	1	0.00165
dfat sp	2.7536	0.066	1	0.79758
chernobyl	3.2003	0.043	1	0.83485
ALL		14.607	5	0.01218

**Figure A.3.8.** Jarque-Bera test in VEC modeling for male lung cancer mortality in Poland

Jarque-Bera test

Equation	chi2	df	Prob > chi2
D_d1mp1r	0.063	2	0.96880
D_d1mp1rd	0.549	2	0.76009
D_smokep1	2.254	2	0.32399
D_dso2p1	8.956	2	0.01136
D_chernoby1	0.103	2	0.94979
ALL	11.925	10	0.29009

Skewness test

Equation	Skewness	chi2	df	Prob > chi2
D_d1mp1r	.05241	0.012	1	0.91312
D_d1mp1rd	.00514	0.000	1	0.99146
D_smokep1	.37634	0.614	1	0.43338
D_dso2p1	-1.048	4.759	1	0.02914
D_chernoby1	-.15337	0.102	1	0.74952
ALL		5.487	5	0.35937

Kurtosis test

Equation	Kurtosis	chi2	df	Prob > chi2
D_d1mp1r	2.782	0.051	1	0.82047
D_d1mp1rd	2.2884	0.549	1	0.45892
D_smokep1	4.2305	1.640	1	0.20028
D_dso2p1	4.9682	4.197	1	0.04050
D_chernoby1	2.9683	0.001	1	0.97372
ALL		6.438	5	0.26588

**Figure A.3.9.** Jarque-Bera test in VEC modeling for female lung cancer mortality in Poland

Jarque-Bera test

Equation	chi2	df	Prob > chi2
D_d1mp1r	0.102	2	0.95019
D_d1mp1rd	0.628	2	0.73045
D_smokep1	0.347	2	0.84062
D_dso2p1	1.673	2	0.43320
D_chernoby1	0.192	2	0.90854
ALL	2.943	10	0.98274

Skewness test

Equation	Skewness	chi2	df	Prob > chi2
D_d1mp1r	.09076	0.036	1	0.85015
D_d1mp1rd	.14864	0.096	1	0.75700
D_smokep1	.06574	0.019	1	0.89115
D_dso2p1	-.55987	1.358	1	0.24383
D_chernoby1	-.21029	0.192	1	0.66156
ALL		1.700	5	0.88889

Kurtosis test

Equation	Kurtosis	chi2	df	Prob > chi2
D_d1mp1r	2.7523	0.066	1	0.79651
D_d1mp1rd	2.2989	0.532	1	0.46558
D_smokep1	3.5507	0.329	1	0.56653
D_dso2p1	3.5391	0.315	1	0.57473
D_chernoby1	2.9864	0.000	1	0.98872
ALL		1.242	5	0.94074

**Figure A.3.10.** Jarque-Bera test in VEC modeling for total lung cancer mortality in Poland

Jarque-Bera test

Equation	chi2	df	Prob > chi2
D_1tp1r	1.799	2	0.40682
D_1tp1rd	58.083	2	0.00000
D_smokep1	0.312	2	0.85575
D_dso2p1	1.323	2	0.51607
D_chernoby1	0.558	2	0.75651
ALL	62.075	10	0.00000

Skewness test

Equation	Skewness	chi2	df	Prob > chi2
D_1tp1r	.55353	1.328	1	0.24922
D_1tp1rd	2.1344	19.741	1	0.00001
D_smokep1	.15341	0.102	1	0.74946
D_dso2p1	.53831	1.256	1	0.26247
D_chernoby1	.00383	0.000	1	0.99364
ALL		22.426	5	0.00043

Kurtosis test

Equation	kurtosis	chi2	df	Prob > chi2
D_1tp1r	3.6594	0.471	1	0.49249
D_1tp1rd	8.9492	38.343	1	0.00000
D_smokep1	3.4398	0.210	1	0.64711
D_dso2p1	2.7507	0.067	1	0.79526
D_chernoby1	2.2823	0.558	1	0.45506
ALL		39.649	5	0.00000

**Figure A.3.11.** Jarque-Bera test in VEC modeling for male lung cancer mortality in Spain

Jarque-Bera test

Equation	chi2	df	Prob > chi2
D_d21mspr	0.366	2	0.83267
D_d21msprd	0.417	2	0.81163
D_smokesp	1.290	2	0.52464
D_dso2sp	1.216	2	0.54447
D_chernoby1	0.389	2	0.82308
ALL	3.679	10	0.96067

Skewness test

Equation	Skewness	chi2	df	Prob > chi2
D_d21mspr	.0518	0.012	1	0.91414
D_d21msprd	-.18705	0.152	1	0.69699
D_smokesp	-.54556	1.290	1	0.25610
D_dso2sp	-.42105	0.768	1	0.38076
D_chernoby1	.28332	0.348	1	0.55534
ALL		2.569	5	0.76606

kurtosis test

Equation	kurtosis	chi2	df	Prob > chi2
D_d21mspr	2.4279	0.355	1	0.55152
D_d21msprd	2.5047	0.266	1	0.60615
D_smokesp	3.018	0.000	1	0.98502
D_dso2sp	3.6428	0.448	1	0.50345
D_chernoby1	2.8042	0.042	1	0.83848
ALL		1.110	5	0.95322

**Figure A.3.12.** Jarque-Bera test in VEC modeling for female lung cancer mortality in Spain

Jarque-Bera test

Equation	chi2	df	Prob > chi2
D_d21fspr	0.502	2	0.77795
D_d21fsprd	0.749	2	0.68774
D_smokesp	0.250	2	0.88257
D_dso2sp	0.250	2	0.88259
D_chernoby1	0.147	2	0.92907
ALL	1.898	10	0.99707

Skewness test

Equation	Skewness	chi2	df	Prob > chi2
D_d21fspr	.03608	0.006	1	0.94014
D_d21fsprd	-.3404	0.502	1	0.47857
D_smokesp	-.22949	0.228	1	0.63284
D_dso2sp	.22084	0.211	1	0.64572
D_chernoby1	-.12917	0.072	1	0.78801
ALL		1.020	5	0.96097

kurtosis test

Equation	kurtosis	chi2	df	Prob > chi2
D_d21fspr	2.323	0.497	1	0.48102
D_d21fsprd	3.4771	0.247	1	0.61950
D_smokesp	3.1412	0.022	1	0.88314
D_dso2sp	2.8116	0.038	1	0.84452
D_chernoby1	2.7372	0.075	1	0.78442
ALL		0.878	5	0.97179

**Figure A.3.13.** Jarque-Bera test in VEC modeling for female lung cancer mortality in Spain

Jarque-Bera test

Equation	chi2	df	Prob > chi2
D_d1tspr	0.057	2	0.97180
D_d1tsprd	0.093	2	0.95480
D_smokesp	1.235	2	0.53919
D_dso2sp	1.207	2	0.54685
D_chernoby1	0.333	2	0.84654
ALL	2.925	10	0.98312

Skewness test

Equation	Skewness	chi2	df	Prob > chi2
D_d1tspr	-.11172	0.054	1	0.81609
D_d1tsprd	-.12632	0.069	1	0.79258
D_smokesp	-.40132	0.698	1	0.40348
D_dso2sp	.5115	1.134	1	0.28698
D_chernoby1	.24001	0.250	1	0.61734
ALL		2.205	5	0.82018

Kurtosis test

Equation	Kurtosis	chi2	df	Prob > chi2
D_d1tspr	2.9464	0.003	1	0.95553
D_d1tsprd	2.8531	0.023	1	0.87851
D_smokesp	2.2956	0.537	1	0.46349
D_dso2sp	3.2603	0.073	1	0.78641
D_chernoby1	2.7222	0.084	1	0.77251
ALL		0.721	5	0.98180

**Table A.4.1.** Additional output for VEC model for total lung cancer Polish mortality

D_1tp1rd						
_ce1						
L1.	.8112038	1.127633	0.72	0.472	-1.398916	3.021324
1tp1r						
LD.	-6.733777	3.691744	-1.82	0.068	-13.96946	.5019075
L2D.	-2.31024	3.293669	-0.70	0.483	-8.765712	4.145232
L3D.	2.094885	3.42556	0.61	0.541	-4.619089	8.808859
1tp1rd						
LD.	6.873984	8.418919	0.82	0.414	-9.626794	23.37476
L2D.	-2.367318	7.381037	-0.32	0.748	-16.83388	12.09925
L3D.	-8.09239	6.620962	-1.22	0.222	-21.06924	4.884456
smokep1						
LD.	8.75e-07	1.09e-06	0.80	0.422	-1.26e-06	3.01e-06
L2D.	7.84e-07	8.57e-07	0.91	0.360	-8.95e-07	2.46e-06
L3D.	4.65e-07	6.23e-07	0.75	0.455	-7.56e-07	1.69e-06
dso2p1						
LD.	-.003101	.0035728	-0.87	0.385	-.0101036	.0039016
L2D.	-.0018456	.0027043	-0.68	0.495	-.007146	.0034548
L3D.	-.0012666	.0014793	-0.86	0.392	-.0041661	.0016328
chernoby1						
LD.	-.3132561	.3520147	-0.89	0.374	-1.003192	.37668
L2D.	.0668705	.2639112	0.25	0.800	-.4503859	.584127
L3D.	.2984761	.2453832	1.22	0.224	-1.1824662	.7794183
_cons	.0042599	.0060433	0.70	0.481	-.0075847	.0161045
D_smokep1						
_ce1						
L1.	499754.7	601885	0.83	0.406	-679918.3	1679428
1tp1r						
LD.	-3027257	1970504	-1.54	0.124	-6889374	834860.3
L2D.	118062.7	1758028	0.07	0.946	-3327608	3563734
L3D.	1815712	1828426	0.99	0.321	-1767936	5399361
1tp1rd						
LD.	3981890	4493680	0.89	0.376	-4825561	1.28e+07
L2D.	-2182506	3939700	-0.55	0.580	-9904177	5539164
L3D.	-6119824	3534003	-1.73	0.083	-1.30e+07	806693.7
smokep1						
LD.	.0219901	.5817318	0.04	0.970	-1.118183	1.162163
L2D.	.2130379	.4572172	0.47	0.641	-.6830914	1.109167
L3D.	-.0095935	.3323759	-0.03	0.977	-.6610382	.6418513
dso2p1						
LD.	-1883.611	1907.029	-0.99	0.323	-5621.318	1854.096
L2D.	-425.7407	1443.461	-0.29	0.768	-3254.873	2403.392
L3D.	-627.5766	789.6119	-0.79	0.427	-2175.188	920.0343
chernoby1						
LD.	-172214.6	187891.3	-0.92	0.359	-540474.7	196045.5
L2D.	67447.65	140865.2	0.48	0.632	-.208643	343538.3
L3D.	220490.6	130975.7	1.68	0.092	-36216.96	477198.3
_cons	.0021903	3225.664	0.00	1.000	-6322.182	6322.187
D_dso2p1						
_ce1						
L1.	-816.7028	287.7839	-2.84	0.005	-1380.749	-252.6566
1tp1r						
LD.	-27.2092	942.1723	-0.03	0.977	-1873.833	1819.415
L2D.	1136.317	840.5794	1.35	0.176	-511.1878	2783.823
L3D.	838.9606	874.2393	0.96	0.337	-874.517	2552.438
1tp1rd						
LD.	4270.572	2148.598	1.99	0.047	59.39698	8481.746
L2D.	256.068	1883.719	0.14	0.892	-3435.954	3948.09
L3D.	234.3458	1689.74	0.14	0.890	-3077.484	3546.175
smokep1						
LD.	-.0007201	.0002781	-2.59	0.010	-.0012653	-.000175
L2D.	-.0002953	.0002186	-1.35	0.177	-.0007237	.0001332
L3D.	-.0001825	.0001589	-1.15	0.251	-.0004939	.000129
dso2p1						
LD.	1.880236	.9118223	2.06	0.039	.0930971	3.667375
L2D.	1.037622	.6901734	1.50	0.133	-.3150931	2.390337
L3D.	.2175993	.3775432	0.58	0.564	-.5223719	.9575704
chernoby1						
LD.	-138.7514	89.8379	-1.54	0.122	-314.8305	37.32762
L2D.	11.50654	67.35295	0.17	0.864	-120.5028	143.5159
L3D.	-11.76788	62.62441	-0.19	0.851	-134.5095	110.9737
_cons	1.342575	1.542311	0.87	0.384	-1.680299	4.36545

**Table A.4.2.** Additional output for VEC model for male lung cancer Polish mortality

D_dso2p1						
_ce1						
L1.	3376.793	2372.445	1.42	0.155	-1273.113	8026.699
_ce2						
L1.	-5559.117	5762.263	-0.96	0.335	-16852.95	5734.711
_ce3						
L1.	.0001797	.0002096	0.86	0.391	-.0002311	.0005906
d1mp1r						
LD.	-3186.871	2319.201	-1.37	0.169	-7732.423	1358.68
L2D.	-2296.924	1963.061	-1.17	0.242	-6144.452	1550.604
L3D.	-1009.322	955.696	-1.06	0.291	-2882.452	863.8074
d1mp1rd						
LD.	5806.971	4363.775	1.33	0.183	-2745.869	14359.81
L2D.	2917.362	3203.565	0.91	0.362	-3361.51	9196.235
L3D.	1192.11	1392.293	0.86	0.392	-1536.734	3920.954
smokep1						
LD.	-.0002763	.0003279	-0.84	0.399	-.000919	.0003664
L2D.	.0003144	.0002054	1.53	0.126	-.0000882	.000717
L3D.	.0001968	.0002142	0.92	0.358	-.000223	.0006166
dso2p1						
LD.	.4439383	.6400807	0.69	0.488	-.8105969	1.698473
L2D.	.1869515	.4313787	0.43	0.665	-.6585352	1.032438
L3D.	.459477	.4556522	1.01	0.313	-.4335848	1.352539
chernoby1						
LD.	.6064695	7.232665	0.08	0.933	-13.56929	14.78223
L2D.	6.509194	5.970391	1.09	0.276	-5.192557	18.21094
L3D.	-4.836119	10.15438	-0.48	0.634	-24.73834	15.0661
_cons	.0544905	4.954767	0.01	0.991	-9.656674	9.765655

**Table A.4.3.** Additional output for VEC model for female lung cancer Polish mortality

D_dso2p1						
_ce1						
L1.	-4377.216	5115.402	-0.86	0.392	-14403.22	5648.788
_ce2						
L1.	12025.99	9101.506	1.32	0.186	-5812.629	29864.62
d1fp1r						
LD.	7988.436	4347.235	1.84	0.066	-531.9888	16508.86
L2D.	869.6059	3241.744	0.27	0.789	-5484.096	7223.308
L3D.	-834.2287	1810.596	-0.46	0.645	-4382.932	2714.475
d1fp1rd						
LD.	-4643.881	8893.166	-0.52	0.602	-22074.17	12786.4
L2D.	-3261.962	6062.518	-0.54	0.591	-15144.28	8620.356
L3D.	6546.69	2906.348	2.25	0.024	850.3524	12243.03
smokep1						
LD.	-.0002302	.000143	-1.61	0.107	-.0005104	.00005
L2D.	.0000856	.0001129	0.76	0.448	-.0001357	.0003068
L3D.	-.0002826	.0001319	-2.14	0.032	-.0005411	-.0000241
dso2p1						
LD.	.3039423	.3886216	0.78	0.434	-.4577419	1.065627
L2D.	-.2817016	.3022482	-0.93	0.351	-.8740972	.310694
L3D.	-.3258061	.2071957	-1.57	0.116	-.7319022	.0802901
chernoby1						
LD.	3.10908	7.174056	0.43	0.665	-10.95181	17.16997
L2D.	5.173901	5.532215	0.94	0.350	-5.669041	16.01684
L3D.	-11.18136	3.945558	-2.83	0.005	-18.91451	-3.448205
_cons	-2.016782	5.288755	-0.38	0.703	-12.38255	8.348987

**Table A.4.4.** Additional output for VEC model for total lung cancer Spanish mortality

<b>D_d1tsprd</b>						
_ce1						
L1.	-1.864358	.8773789	-2.12	0.034	-3.583989	-.144727
d1tspr						
LD.	1.692986	.8614839	1.97	0.049	.0045083	3.381463
L2D.	1.266714	.7200223	1.76	0.079	-.1445037	2.677932
L3D.	.3746541	.3882013	0.97	0.334	-.3862065	1.135515
d1tsprd						
LD.	-2.241319	.9268634	-2.42	0.016	-4.057937	-.4246997
L2D.	-1.90139	.8422591	-2.26	0.024	-3.552187	-.2505922
L3D.	-.414778	.5456809	-0.76	0.447	-1.484293	.6547369
smokesp						
LD.	-6.94e-08	4.62e-08	-1.50	0.133	-1.60e-07	2.11e-08
L2D.	-8.94e-08	4.26e-08	-2.10	0.036	-1.73e-07	-5.86e-09
L3D.	-6.79e-09	1.75e-08	-0.39	0.698	-4.11e-08	2.75e-08
dso2sp						
LD.	-.0003742	.0001645	-2.27	0.023	-.0006967	-.0000517
L2D.	-.0002176	.0001189	-1.83	0.067	-.0004505	.0000154
L3D.	-.0000274	.0000747	-0.37	0.714	-.0001738	.000119
chernoby1						
LD.	.0077651	.0037302	2.08	0.037	.000454	.0150761
L2D.	.0002338	.0040677	0.06	0.954	-.0077387	.0082063
L3D.	.0070493	.0041566	1.70	0.090	-.0010975	.015196
_cons	.0001917	.0002065	0.93	0.353	-.0002129	.0005963
<b>D_smokesp</b>						
_ce1						
L1.	3.49e+07	1.78e+07	1.96	0.050	-58081.32	6.98e+07
d1tspr						
LD.	-2.64e+07	1.75e+07	-1.51	0.132	-6.07e+07	7929845
L2D.	-1.92e+07	1.46e+07	-1.31	0.190	-4.78e+07	9514970
L3D.	-3739290	7885755	-0.47	0.635	-1.92e+07	1.17e+07
d1tsprd						
LD.	2.63e+07	1.88e+07	1.40	0.163	-1.06e+07	6.32e+07
L2D.	2.35e+07	1.71e+07	1.37	0.170	-1.00e+07	5.70e+07
L3D.	1.21e+07	1.11e+07	1.09	0.276	-9640991	3.38e+07
smokesp						
LD.	.7638294	.9375067	0.81	0.415	-1.07365	2.601309
L2D.	.7238469	.8660237	0.84	0.403	-.9735283	2.421222
L3D.	-.55712	.3556192	-1.57	0.117	-1.254121	.1398808
dso2sp						
LD.	7891.254	3342.544	2.36	0.018	1339.988	14442.52
L2D.	6430.491	2414.487	2.66	0.008	1698.183	11162.8
L3D.	2222.53	1517.331	1.46	0.143	-751.3842	5196.443
chernoby1						
LD.	-58306.43	75773.85	-0.77	0.442	-206820.4	90207.57
L2D.	-47802.57	82629.3	-0.58	0.563	-209753	114147.9
L3D.	-135658	84435.02	-1.61	0.108	-301147.6	29831.58
_cons	6.12e-06	4193.811	0.00	1.000	-8219.719	8219.719
<b>D_dso2sp</b>						
_ce1						
L1.	839.9558	4283.736	0.20	0.845	-7556.013	9235.924
d1tspr						
LD.	339.0693	4206.13	0.08	0.936	-7904.793	8582.932
L2D.	575.1182	3515.454	0.16	0.870	-6315.045	7465.282
L3D.	450.7302	1895.363	0.24	0.812	-3264.114	4165.574
d1tsprd						
LD.	-1631.775	4525.34	-0.36	0.718	-10501.28	7237.729
L2D.	-770.062	4112.266	-0.19	0.851	-8829.955	7289.831
L3D.	-714.4289	2664.246	-0.27	0.789	-5936.254	4507.396
smokesp						
LD.	-.0000243	.0002253	-0.11	0.914	-.0004659	.0004173
L2D.	-.0000204	.0002082	-0.10	0.922	-.0004283	.0003876
L3D.	-.0000182	.0000855	-0.21	0.831	-.0001857	.0001493
dso2sp						
LD.	-.5210295	.8033898	-0.65	0.517	-2.095645	1.053586
L2D.	-.2937203	.5803288	-0.51	0.613	-1.431144	.8437032
L3D.	-.0106107	.3646947	-0.03	0.977	-.7253992	.7041779
chernoby1						
LD.	-2.601041	18.21246	-0.14	0.886	-38.2968	33.09472
L2D.	-2.163061	19.86018	-0.11	0.913	-41.0883	36.76218
L3D.	11.80746	20.29419	0.58	0.561	-27.96843	51.58335
_cons	-.2164088	1.007994	-0.21	0.830	-2.192041	1.759224
<b>D_chernoby1</b>						
_ce1						
L1.	-597.849	185.3881	-3.22	0.001	-961.203	-234.4949
d1tspr						
LD.	457.9239	182.0295	2.52	0.012	101.1526	814.6953
L2D.	322.6634	152.139	2.12	0.034	24.47638	620.8505
L3D.	66.6961	82.02604	0.81	0.416	-94.07198	227.4642
d1tsprd						
LD.	-472.8373	195.8441	-2.41	0.016	-856.6847	-88.98997
L2D.	-403.4283	177.9674	-2.27	0.023	-752.238	-54.61867
L3D.	-182.0783	115.3011	-1.58	0.114	-408.0643	43.90776
smokesp						
LD.	-.0000163	9.75e-06	-1.67	0.094	-.0000354	2.79e-06
L2D.	-.0000195	9.01e-06	-2.16	0.031	-.0000371	-1.81e-06
L3D.	5.22e-06	3.70e-06	1.41	0.158	-2.03e-06	.0000125
dso2sp						
LD.	-.1220092	.0347685	-3.51	0.000	-.1901542	-.0538643
L2D.	-.0973987	.025115	-3.88	0.000	-.1466232	-.0481742
L3D.	-.0448353	.015783	-2.84	0.005	-.0757694	-.0139013
chernoby1						
LD.	1.986782	.7881843	2.52	0.012	.4419689	3.531595
L2D.	.5295833	.8594933	0.62	0.538	-1.154993	2.214159
L3D.	2.359481	.8782761	2.69	0.007	.6380914	4.08087

**Table A.4.5.** Additional output for VEC model for male lung cancer Spanish mortality

D_dso2sp						
_ce1						
L1.	3492.971	3214.682	1.09	0.277	-2807.689	9793.631
_ce2						
L1.	-5545.071	3255.284	-1.70	0.088	-11925.31	835.1684
d21mspr						
LD.	-2607.87	2780.263	-0.94	0.348	-8057.085	2841.345
L2D.	-1330.836	1802.102	-0.74	0.460	-4862.891	2201.219
L3D.	-295.7644	721.2647	-0.41	0.682	-1709.417	1117.888
d21msprd						
LD.	4397.375	2852.043	1.54	0.123	-1192.526	9987.277
L2D.	2674.905	1946.512	1.37	0.169	-1140.188	6489.997
L3D.	1252.566	906.8022	1.38	0.167	-524.734	3029.865
smokesp						
LD.	-.0000886	.0001589	-0.56	0.577	-.0004001	.000223
L2D.	-.0001036	.0001332	-0.78	0.437	-.0003647	.0001575
L3D.	.0000202	.0000817	0.25	0.805	-.0001399	.0001803
dso2sp						
LD.	-.3355713	.678547	-0.49	0.621	-1.665499	.9943564
L2D.	-.091762	.495481	-0.19	0.853	-1.062887	.879363
L3D.	-.0950523	.3511051	-0.27	0.787	-.7832056	.593101
chernoby1						
LD.	-2.288044	13.87189	-0.16	0.869	-29.47645	24.90036
L2D.	1.746658	18.91169	0.09	0.926	-35.31957	38.81289
L3D.	26.80115	15.46126	1.73	0.083	-3.502369	57.10468
_cons	-.0226934	8.381792	-0.00	0.998	-16.4507	16.40532