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Regression Analysis of Coal Production and Tobacco Smoking with Lung Cancer in West Virginia, USA

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Abstract

The objective of this study is to investigate the impacts of coal production on the prevalence of lung cancer in West Virginia, USA. It has been accepted that lung cancer is strongly correlated with tobacco smoking. In this study, we used linear regression analysis to estimate the degree of association of lung cancer incidence and tobacco smoking prevalence in the presence of coal production. We found that lung cancer association with tobacco smoking becomes stronger when coal production is considered in the producing coal counties. In coal non-producing counties, the association between lung cancer and tobacco smoking is not significant due to high variability of lung cancer incidence and limited availability of data. When we used a first order inverse distance to estimate the effective coal production in coal non-producing counties, we found a moderate correlation between lung cancer and tobacco smoking. The study results demonstrated that effects of coal production potentiate the association between tobacco smoking on lung cancer development.

Keywords: Coal production • Lung cancer • Tobacco smoking • Linear regression analysis • GPS mapping

Introduction

Lung cancer is a devastating disease that causes progressive breathing difficulty and pain to the patient [1,2]. Lung cancer complications often make the management of the disease very costly and intolerable [3]. Lung cancer not only affects the function of the lung but compromises the function of other organs, chiefly the heart and lead cardiovascular disease [4]. Patient habits and environmental causes often accelerate the effects of lung cancer [5]. In 2020, it is estimated that new cases of 197,453 and 136,084 Americans die because of lung cancer [6]. According to the 2021 National Institute of Health annual report the cost related to lung cancer health care is estimated to be 1.35 billion in 2019 [7]. Although lung cancer death has decreased the cost of management of patients with lung cancer is increasing. In particular, in West Virginia the rate of deaths from lung cancer is estimated at 1,339 deaths with 1,875 new cases, second only to Kentucky. The death from lung cancer is the highest in Kentucky followed by West Virginia. Although tobacco smoking has been identified as the primary cause of lung cancer, environmental and industrial pollutants seem to contribute to the statistics of lung cancer.

West Virginia economy relies mostly on coal industry for electric power. In fact, about 91% of electric power in WV comes from coal-fired electric power [8]. Although coal energy is in accelerating decline the effects of coal mining is yet to be determined [9]. Coal is classified as a nonrenewable energy extracted from sedimentary deposits. Coal consists mostly of carbon and hydrocarbon [10]. But it also contains other elements such as hydrogen, sulfur, oxygen and nitrogen in small amounts. Some of the largest coal deposits are located in many fields throughout the mid and western regions of the US. The biggest coal deposit by volume is in Wyoming and Montana. Large coal deposits can

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Received: 01 November, 2023, Manuscript No. IJPHS-23-119920; **Editor Assigned:** 03 November, 2023, PreQC No. P-119920; **Reviewed:** 15 November, 2023, QC No. Q-119920; **Revised:** 21 November, 2023, Manuscript No. R-119920; **Published:** 29 November, 2023, DOI: 10.37421/2736-6189.2023.8.354 also be found in North Dakota, Montana, Ohio, Pennsylvania, West Virginia, Virginia, Alabama, Illinois and Indiana. Although coal is used in many industries, it is predominately used for electric power generation. It also can be used in cement, carbon fibers, foams, medicines, synthetic fuels and heating. The heat generated by the burning of coal turns turbines to transform mechanical energy to electric energy [11]. In 2021, over 22% of all electricity was produced from coal energy [12]. The efficiency of directly burning coal is less than 50% and depends on the pre- processing of coal and the plant itself [13]. Expensive integrated gasification combined cycle power plants gasify coal than burn it more efficiently to produce electricity and thus emit less pollutants [14].

The burning of coal generates burnt coal that contains moisture and millions of tons of ash, sulfur, phosphorus and many toxic gases and minerals [15,16]. The coal industry pollutes air and water. Furthermore, coal burning byproducts are often warehoused near the coal plants. Coal oxides byproducts directly affect the acidity of water systems and groundwater [17]. Consequently, coal mining, transporting, burning and storing cause major damage to human health, water resources, wild life and the environment [18]. Coal burning produces Carbon Dioxide (CO₂) and Sulfur Dioxide (SO₂). SO₂ not only precipitates into acid rain and worsens the acidity of water resources, but also contributes to acidification of ecosystems. SO₂ creates matter particulates of size 2.5 mm that invades human tissues and makes air pollution more dangerous. CO₂ remains in the atmosphere as a greenhouse gas and contributes substantially to the worsening climate change [19].

Globally, untimely deaths and acute and chronic diseases are brought on by the mining and usage of coal [20,21]. More human deaths are even higher around coal power plants [22]. Coal plants pollutants trigger asthma, exasperate chronic obstructive pulmonary diseases and cause lung cancer [23]. Breathing coal dust causes pneumoconiosis or black lung in coal miners [24]. The leading cause of cancer-related fatalities globally is lung cancer, which is also the most often diagnosed cancer [25]. There are several types of lung cancer that preferentially attack the various cells of the lining or parenchymal tissues of the lungs [26]. Among the types of lung cancer, squamous cell carcinoma is highly associated with tobacco smoking and secondhand smoke [27]. Tobacco smoking contains carcinogens that cause immediate changes in the lung tissues [28]. It is believed that when the cells of the lungs are repeatedly exposed to smoking, it causes them to transform to cancerous cells that are more resilient. The risk factors of lung cancer include smoking, exposure to secondhand smoke, previous radiation therapy, exposure to asbestos and other carcinogens, dietary habits and family history

of lung cancer [29-31]. In addition, lung cancer has been linked to continuous coal burning and ash and wood burning [32].

The purpose of this study is to investigate the impacts of coal production on the prevalence of lung cancer in West Virginia, USA. Linear regression analysis was used to estimate the degree of association of lung cancer incidence and tobacco smoking prevalence in the presence of coal production. Furthermore, a first order inverse distance was used to estimate the effective coal production in coal non-producing counties.

Methods

In this study we researched the association between lung cancer and coal production in the counties of West Virginia using regressing models adjusted to smoking as a confounding variable from the year 2012 through 2016 using federal, state and private data registries.

Study design

This study consisted of a retrospective analysis of data purged from federal and state agencies from 2012 to 2016. We used data from NIH Cancer Institute, West Virginia Department of Health, West Virginia Coal Association and West Virginia Office of Commerce resources and publications for this study. The NIH Cancer Institute maintains cancer registry for the US population. The West Virginia Department Health maintains specific data registry on burden and cost of cancer in the state of West Virginia. Both registries represent most cancers affecting the population of West Virginia per county. They also include selected socioeconomic characteristics of patients with cancer. In this study all patients of the state of West Virginia with lung cancer and bronchus from the year 2012 through 2016 are included. We used the West Virginia Department of Health to obtain the tobacco smoking prevalence per county from 2012 through 2016. We used the West Virginia Coal Association and the West Virginia Office of Department of Commerce capture yearly coal production per county between the year 2012 and 2016. Coal production represents the total production of ground and above ground mines.

Statistical analysis

We used descriptive statistics to evaluate the center of tendency and variability in the data. We assumed the data to be normally distributed. The association among lung cancer, smoking and coal production is estimated through multivariate regression models. In these regression models the outcome of lung cancer associations were separately performed to coal production and smoking rate and in combination to adjust for confounding effect of smoking. These models were applied in three cases of data stratifications: Group 1, Group 2 and Group 3. Group 1 consisted of coal producing counties. Group 2 consisted of coal non-producing counties. Group 3 consisted of all counties with effective-producing counties, we used the inverse distance method. For the coal non-producing counties, we used the inverse distance method to estimate the effective coal production that may have an impact on lung cancer incidence in the corresponding county. The implementation of the linear regression and inverse distance methods were detailed in the Data Mapping and Processing Section.

The linear regression analysis of lung cancer, coal production and smoking was performed in the three groups, where lung cancer incidence is the dependent variable and coal production and tobacco smoking are the independent variables. The p-values less than 0.05 were two-sided and considered statistically significant. We used Matlab R2023a (MathWorks, Inc., Natick, Massachusetts, USA) for data processing and statistical analysis.

Data mapping and processing

Prior to regression analysis, we normalized and standardized data to allow for comparison between the regression coefficients of the three groups. While normalization maintains a consistent format for the data, standardization assumes that the data is normally distributed [33,34]. Although health and social data may not be normally distributed, we visually inspected the residuals of the regression analysis to be either normal or near normal distributions around zero mean and quasi- constant variance [35]. We used multivariate linear regression models to estimate the degree of association between lung cancer, coal production and tobacco smoking. The following expression represents the linear regression model:

$$Y = b_0 + b_1 X_1 + b_2 X_2 \quad (1)$$

Y represents lung cancer incidence (dependent variable). X1 and X2 represent coal production (Tons) and tobacco smoking prevalence (independent variables) respectively. The parameters bi are the estimated regression coefficients and represent the change of lung cancer relative to a one- unit change in one independent variable while the other independent variable is kept constant at the same value.

To determine the degree of association between lung cancer and coal production and adjusted the confounding variable of tobacco smoking, we designed three data stratification cases. We designed the following data groups. The first group consists of the coal producing counties and ignoring the non-producing counties. In this group we only analyze the association between lung cancer and coal production and smoking in the coal producing counties. The second group consists of the non- producing coal counties only. In this group we analyze the association between long cancer and smoking. The third group consists of all counties. We used the inverse distance principle to estimate an effective (apparent) coal production amount. Not all counties produce coal. Actually, most counties do not produce coal for the lack of coal mines. The locations of all counties are geographically represented by the Global Positioning System (GPS). The inverse distance method is used to estimate data points based on weighted average using known data points. The inverse weighted distance expression is described by the following expressions as:

$$u(x) = \begin{cases} \frac{\sum_{i=0}^{N} W_i(x) u_i}{\sum_{i=1}^{N} W_i(x)}, & \text{if } d(x, x_i) \neq 0 \text{ for all } i, \\ u_i & \text{if } d(x, x_i) = 0 \text{ for some } i, \end{cases}$$
(2a)

The inverse distance weight is described by the following expression as:

$$v_i(x) = \frac{1}{d(x, x_i)^p}$$
 (2b)

The effective total coal production for non-producing counties is estimated using the inverse distance weighting method for p=1. Thus, the effective coal production of a non-producing county is the amount of coal produced by the averaging principle, although there is no actual coal production in that county. We used the following expression to estimate the distance between each two counties, in miles, as:

$$\begin{array}{l} (i, j) = 3963 |\cos^{-1}[(\sin(x_i) \sin(x_j) + \cos(x_i) \cos(x_j) \cos(y_i - y_j)]| \\ for \quad i \neq j \end{array} \tag{3}$$

The variables \mathbf{x}_i and \mathbf{y}_i are the latitudes and longitudes data for each county in West Virginia.

Results and Discussion

Table 1 lists the mean and standard deviation values of coal production, tobacco prevalence, lung cancer incidence in the state of West Virginia for years 2012 to 2016. The descriptive statistics are listed for the coal producing and non-producing counties. For coal producing counties, coal production mean ranges from 4.0 to 5.0 million short tons with a standard deviation ranging from 4.4 to 5.0 million short tons. During the years of 2012 to 2016, West Virginia's coal production represented over 13% of the total coal production of the USA and showed high variability due to changing demand, technology, competing energy resources, public concerns and workforce. Nonetheless, West Virginia's coal production declined in 2016 by 22.5% as compared to its highest in 2014. In 2016, fewer counties produced coal. The mean and standard deviation values of tobacco prevalence and lung cancer incidence ranged from 0.357 (SD 0.050) to 0.360 (SD 0.056) and 82.89 (SD 16.43)

Table 1. Coal production, tobacco prevalence and lung cancer incidence mean and standard deviation values for years 2012 through 2016 in coal producing and coal non-producing counties of West Virginia, USA.

Year	Coal Production (Short tones × 10 ⁶)	Tobacco Prevalence (%)	Lung Cancer Incidence Rate (per 100,000)
2012 - (N _{pc} = 28)	4.626 ± 4.997	0.357 ± 0.050	82.89 ± 16.43
2013 - (N _{pc} = 25)	4.814 ± 4.674	0.358 ± 0.053	83.92 ± 15.48
2014 - (N _{pc} = 25)	4.942 ± 4.878	0.358 ± 0.053	83.68 ± 15.75
2015 - (N _{pc} = 25)	4.118 ± 4.407	0.357 ± 0.054	83.06 ± 17.14
2016 - (N _{pc} = 22)	4.019 ± 4.460	0.360 ± 0.056	83.95 ± 17.11
2012 - (N _{npc} =27)	-	0.340 ± 0.051	75.69 ± 11.74
2013 - (N _{npc} =30)	-	0.340 ± 0.048	75.56 ± 12.96
2014 - (N _{npc} =30)	-	0.340 ± 0.048	75.52 ± 12.82
2015 - (N _{npc} =30)	-	0.340 ± 0.048	76.30 ± 12.67
2016 - (N _{npc} = 33)	-	0.340 ± 0.046	75.74 ± 14.26

to 83.95 (SD 17.11) respectively in the coal producing counties, whereas those in the coal non-producing counties ranged from 0.340 (SD 0.046) to 0.340 (SD 0.051) and 75.52 (SD 12.82) to 76.30 respectively. Lung cancer incidence mean values are about 10% less in the coal non-producing counties as compared to those in coal producing counties. However, they are still more than the average nationwide indicating coal production may play a role in the incidence of lung cancer.

Figures 1-3 represent illustrations of colour mapped gradient of coal production, lung cancer incidence and tobacco smoking rate respectively for all counties of West Virginia during 2016. A quantitative association between coal production and lung cancer incidence may be seen in Figures 1 and 2. Figure 4 shows the changes of tobacco smoking rate and lung cancer incidence as functions of increasing coal production in coal producing counties only. Table 2 lists the results of the linear regression analysis of normalized coal production, tobacco prevalence and lung cancer incidence data for the years 2012 to 2016. When only coal production is considered as the independent variable, the R² and adjusted R² values are less than 0.1 with (p>0.05) indicating that the association between coal production and lung cancer incidence is weak and without no statistical significance. Without considering other variables, coal production does not seem to be a risk for lung cancer incidence. The R² and adjusted R² values are low indicating that the correlation between lung cancer incidence and coal production is poor due to collinearity between the data vectors. We attribute the low values of R² not only to the factors associated with coal production but also to the limited data available in state and federal registries. The factors associated with coal production include transportation, dust, climate conditions, particulate size and coal burning by-products. The negative adjusted R² values indicate that these factors may be independent and have different impacts on lung cancer incidence.

When tobacco smoking rate is considered as the only independent variable, it is strongly correlated to lung cancer incidence with a coefficient of correlation ranging from 0.630 to 0.702 and R² and adjusted R² values ranging from 0.276 to 0.287 and 0.245 to 0.257 respectively (p<0.05). This study corroborates the association between lung cancer incidence and tobacco smoking rate. When the linear regression analysis is implemented with both coal production and tobacco prevalence as independent variables and lung cancer incidence as the dependent variable, the R^2 and adjusted R^2 values significantly (p<0.05) increased from 0.293 to 0.453 and 0.218 to 0.409 respectively. Similarly, the coefficient of correlation between coal production and lung cancer in the presence of tobacco smoking increased significantly during the years 2012 to 2014, but for the years 2015 and 2016 did not showed a statistical significance perhaps due to decreased coal production. Table 3 lists the results of the linear regression analysis of standardized coal production, tobacco prevalence and lung cancer incidence data for the years 2012 to 2016. Similar patterns were observed substantiating the results of the normalization method.

Tables 4 and 5 list the linear regression analysis of normalized and standardized tobacco smoking prevalence and lung cancer incidence in the coal non-producing counties. No statistical significance was found between tobacco smoking prevalence and lung cancer incidence. The R^2 and adjusted R^2 values were less than 0.02 (with the exception of 2016) suggesting the lack of correlation between tobacco smoking and lung cancer contrary to common findings. The lack of statistical significance (p>0.05) is, perhaps, due to missing data and high variability with respect to the mean as illustrated in Table 1. The lack of statistical significance does not remove the risk of tobacco



Figure 1. Illustration of coal production (million short tons) per county in West Virginia, USA, during 2016.



Figure 2. Illustration of lung cancer incidence (per 100,000) per county in West Virginia, USA, during 2016.

smoking on lung cancer. We expected that the high variability in lung cancer incidence and the small size of data played a role in the lack of significance in the results of the regression analysis. Not surprisingly, the lack of statistical significance may actually substantiate the potentiating role of coal production on lung cancer incidence even in the coal non-producing counties as well. It is clear that, although the coal non-producing counties do not produce coal, the effects of coal production in the nearby counties may have played a significant



Figure 3. Illustration of tobacco smoking rate (percentage) per county in West Virginia, USA, during 2016.



Figure 4. Changes of Tobacco Smoking Rate (TSR) and Lung Cancer Incidence (LCI) as functions of increasing (real) Coal Production (CP) for years 2012 through 2016 in coal producing counties of West Virginia, USA.

role that needs to be estimated from the coal producing counties.

The linear regression analysis of the producing counties shows that tobacco smoking prevalence is moderately associated with lung cancer incidence when considered alone as an independent variable. That correlation increased significantly when both coal production and tobacco smoking prevalence are included in the regression analysis as independent variables, indicating that there is an added effect from coal production. We expanded the regression analysis to all counties using a first order distance method to estimate an effective coal production in coal non-producing counties. The effective coal production is not real, but rather it is an amount of coal that would have been produced by a coal non-producing county using the nearby coal producing counties. Figure 5 shows the changes of tobacco smoking prevalence and lung cancer incidence as functions of increasing coal production (real and effective) in all counties. Figures 4 and 5 were plotted as scattered diagrams to illustrate the degree of linearity of lung cancer incidence, tobacco smoking and coal

Table 2. Linear regression analysis results among normalized (real) coal production, tobacco smoking rate, and lung cancer prevalence data for years 2012 through 2016 in coal producing counties of West Virginia, USA.

	Lung Cancer	Coefficient(s)	P value	RMS-error	R ²	Adjusted R ²
	Coal	0.358	0.051	0.265	0.138	0.105
2012 -	Tobacco	0.688	0.004	0.242	0.285	0.257
	Ocal Tabaaaa	0.396	0.01	0.215	0.453	0.409
	Cual, Tubaccu	0.725	0.001	-	-	-
	Coal	0.271	0.224	0.287	0.064	0.023
0010	Tobacco	0.673	0.007	0.253	0.276	0.245
2013 -	Oral Tabaaaa	0.391	0.04	0.234	0.404	0.35
	Coal, lobacco	0.761	0.002	-	-	-
2014 -	Coal	0.122	0.581	0.307	0.013	-0.03
	Tobacco	0.702	0.007	0.263	0.277	0.245
	Coal, Tobacco	0.3	0.124	0.255	0.352	0.293
		0.809	0.003	-	-	-
	Coal	0.048	0.822	0.277	0.002	-0.041
0015	Tobacco	0.63	0.006	0.235	0.284	0.252
2015 -	Ocal Tabaaaa	0.207	0.271	0.233	0.323	0.261
	Coal, Tobacco	0.696	0.004	-	-	-
	Coal	-0.156	0.543	0.307	0.019	-0.03
-	Tobacco	0.67	0.01	0.262	0.287	0.251
2017 -	Ocal Tabaaaa	0.098	0.687	0.267	0.293	0.218
	Coal, TODACCO	0.712	0.014	-	-	-

Table 3. Linear regression analysis results among standardized (real) coal production, tobacco smoking rate, and lung cancer prevalence data for years 2012 through 2016 in coal producing counties of West Virginia, USA.

	Lung Cancer	Coefficient(s)	P value	RMS-error	R ²	Adjusted R ²
	Coal	0.023	0.051	0.058	0.138	0.105
2012	Tobacco	0.033	0.004	0.053	0.285	0.257
2012	Coal,	0.025	0.01	0.047	0.453	0.409
	Tobacco	0.034	0.001	-	-	-
	Coal	0.016	0.224	0.287	0.064	0.023
2013	Tobacco	0.034	0.007	0.056	0.276	0.245
2013	Coal,	0.024	0.041	0.234	0.404	0.35
	Tobacco	0.038	0.002	-	-	-
	Coal	0.007	0.581	0.064	0.013	-0.03
2014	Tobacco	0.033	0.007	0.055	0.277	0.245
2014	Coal, Tobacco	0.018	0.124	0.053	0.352	0.293
		0.039	0.003	-	-	-
	Coal	0.003	0.822	0.06	0.002	-0.041
2015	Tobacco	0.031	0.006	0.05	0.284	0.252
2010	Coal,	0.012	0.271	0.05	0.323	0.261
	Tobacco	0.034	0.004	-	-	-
	Coal	-0.008	0.543	0.059	0.019	-0.03
2016	Tobacco	0.031	0.01	0.051	0.287	0.251
2010	Coal,	0.005	0.687	0.052	0.293	0.218
	Tobacco	0.033	0.014	-	-	-

production as functions of increasing coal production further strengthening the normality of data.

Table 6 lists the results of the linear regression analysis of normalized coal production (real and effective), tobacco prevalence and lung cancer incidence data for the years 2012 to 2016 for all counties. Except for the years 2012 and 2013, coal production (real and effective) still did not show a statistical association with lung cancer incidence when considered as the

Table 4. Linear regression analysis results between normalized tobacco smoking rate and lung cancer prevalence data for years 2012 through 2016 in coal non-producing counties of West Virginia, USA.

Lung Cancer vs. Tobacco	Coefficient(s)	P value	RMS-error	R ²	Adjusted R ²
2012 - (N _{npc} =27)	0.058	0.795	0.247	0.003	-0.037
2013 - (N _{npc} =30)	0.137	0.526	0.245	0.015	0.021
2014 - (N _{npc} =30)	0.111	0.597	0.238	0.01	-0.025
2015 - (N _{npc} =30)	0.071	0.741	0.242	0.004	-0.032
2016 - (N _{npc} =33)	0.67	0.01	0.262	0.287	0.251

Table 5. Linear regression analysis results between standardized tobacco smoking rate and lung cancer prevalence data for years 2012 through 2016 in coal non-producing counties of West Virginia, USA.

Lung Cancer vs. Tobacco	Coefficient(s)	P value	RMS-error	\mathbb{R}^2	Adjusted R ²
2012 - (N _{npc} = 27)	4.46e-03	0.795	0.087	0.003	-0.037
2013 - (N _{npc} = 30)	9.29e-03	0.526	0.078	0.015	0.021
$2014 - (N_{npc} = 30)$	7.85e-03	0.597	0.079	0.01	-0.025
2015 - (N _{npc} = 30)	4.56e-03	0.741	0.08	0.004	-0.032
2016 - (N _{npc} = 33)	3.13e-03	0.01	0.051	0.287	0.251



Figure 5. Changes of Tobacco Smoking Rate (TSR) and Lung Cancer Incidence (LCI) as functions of increasing (real and effective) Coal Production (CP) for years 2012 through 2016 in all counties of West Virginia, USA.

 Table 6. Linear regression analysis results among normalized (real and effective) coal production, tobacco smoking rate, and lung cancer prevalence data for years 2012 through 2016 in all counties of West Virginia, USA.

	Lung Cancer (all)	Coefficient(s)	P value	RMS-error	R ²	Adjusted R ²
	Coal	0.371	0.008	0.217	0.127	0.11
2012	Tobacco	0.39	0.006	0.216	0.135	0.118
2012	Coal, Tobacco	0.361	0.006	0.202	0.255	0.227
		0.38	0.004	-	-	-
2012	Coal	0.32	0.039	0.223	0.078	0.061
	Tobacco	0.39	0.006	0.216	0.135	0.118
2010	Coal, Tobacco	0.334	0.021	0.208	0.213	0.183
		0.4	0.003	-	-	-
	Coal	0.203	0.172	0.217	0.035	0.017
2014 -	Tobacco	0.384	0.007	0.217	0.13	0.113
	Coal Tobacco	0.241	0.085	0.213	0.179	0.147
	Cuai, 10Dacco	0.41	0.004	-	-	-

2015 -	Coal	0.135	0.41	0.233	0.013	-0.006
	Tobacco	0.373	0.009	0.22	0.121	0.104
	Coal, Tobacco —	0.181	0.244	0.219	0.144	0.119
		0.391	0.007	-	-	-
2016 -	Coal	0.024	0.889	0.231	3.70E-04	-0.019
	Tobacco	0.358	0.012	0.218	0.114	0.098
	Coal Tobacco	0.097	0.558	0.219	0.12	0.086
	Cuai, IUDaucu —	0.371	0.011	-	-	-

 Table 7. Linear regression analysis results among standardized (real and effective)

 coal production, tobacco smoking rate, and lung cancer prevalence data for years 2012

 through 2016 in all counties of West Virginia, USA.

	Lung Cancer (all)	Coefficient(s)	P value	RMS- error	R ²	Adjusted R ²
	Coal	0.356	0.008	0.946	0.127	0.11
2012	Tobacco	0.367	0.006	0.939	0.135	0.118
2012	Coal Tobacco	0.347	0.006	0.88	0.255	0.226
		0.366	0.004	-	-	-
	Coal	0.259	0.039	0.969	0.078	0.061
2013	Tobacco	0.367	0.006	0.939	0.135	0.118
2013 -	Coal, Tobacco	0.291	0.021	0.9	0.219	0.189
		0.376	0.003	-	-	-
2014 -	Coal	0.187	0.172	0.992	0.035	0.016
	Tobacco	0.36	0.007	0.942	0.13	0.113
	Coal, Tobacco	0.222	0.085	0.924	0.179	0.147
		0.381	0.004	-	-	-
	Coal	0.113	0.41	1	0.013	-0.006
2015	Tobacco	0.348	0.009	0.946	0.121	0.104
2010	Coal Tobacco	0.152	0.244	0.943	0.144	0.111
	Coal, Tobacco	0.364	0.007	-	-	-
	Coal	0.019	0.889	1.01	3.70E-04	-0.019
2016	Tobacco	0.338	0.012	0.95	0.114	0.098
2010	Coal Tobacco	0.078	0.558	0.956	0.12	0.086
	Coal, TODACCO	0.351	0.01	-	-	-

only independent variable. The corresponding R² and adjusted R² values were less than 0.13. Tobacco prevalence is somewhat correlated to lung cancer

incidence with a coefficient of correlation ranging from 0.358 to 0.390 and R^2 and adjusted R^2 values ranging from 0.114 to 0.135 and 0.098 to 0.118 respectively (p<0.05). The correlation between tobacco smoking prevalence and lung cancer increased slightly when coal production (real and effective) was added as an independent variable (Table 7). Although the association between tobacco smoking prevalence and lung cancer incidence was lower than of that of coal producing counties nonetheless it demonstrated that lung cancer incidence in coal non-producing counties may have been affected by coal by-products that may have reached them. The first order inverse distance, used in this study, suggests that the effects of coal production decreased by 90% at 10 miles away. However, recently studies have demonstrated that coal dust may travel more than 30 miles under normal conditions. Nonetheless, the results of the linear regression analysis demonstrated the potentiating effect of coal production on tobacco smoking and increase the risk of developing lung cancer in West Virginia.

Conclusion

We believe that this study is the first in looking into the combined association of coal production and tobacco smoking prevalence with lung cancer incidence in West Virginia. We implemented linear regression analysis to estimate the degree of association of lung cancer incidence and tobacco smoking prevalence in the presence of coal production. We also designed averaging methods to estimate the effects of coal production from coal producing counties on coal non- producing counties. We found that lung cancer association with tobacco smoking becomes stronger when coal production is considered in the producing coal counties. In coal non-producing counties, the association between lung cancer and tobacco smoking is not significant due to high variability of lung cancer incidence in limited data. When we used a first order inverse distance to estimate the effective coal production in coal nonproducing counties, we found a moderate correlation between lung cancer and tobacco smoking. We recognize that R² and adjust R² values are low between lung cancer incidence and coal production. We attribute this fact not only to the non-accounted factors associated with goal production but also to the limited data available in state and federal registries. Nonetheless, despite the limited data, this study has demonstrated the potentiating effect of coal production and tobaccos smoking on lung cancer development. The limited energy resources of West Virginia, coal burning will continue to be the prime source of electricity. Our future work is to design controlled studies to mitigate the effects of coal production factors and coal mining on lung cancer incidence.

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Conflict of Interest

The authors have no conflict of interest to disclose.

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