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A lung cancer disease due to smoker using data mining classification algorithm for the prediction

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ABSTRACT

The smoke exposure detection risk comprises of determining the relative risk (RR) of lung cancer in life span by no means smokers linked with environmental tobacco smoke (ETS) disclosure. The most important purpose is to nearby an updated meta-psychiatry of the epidemiological information, even though for a short time converse the tentative confirmation, and learning of smolder forbid. This entire thing aimed to construct an experienced catalog most intimately correspondent to given that spousal smolder is the conventional catalog for learning ETS possessions, women connubial to a smoker having a patently elevated ETS disclosure, as considered by conine, than women connubial to a non-smoker. Thus, the fallout was selected in the subsequent categorize of fondness for Exposure, disease category, a moment of exposure, disease characterization, and label of no ETS revelation. In adding up, consequences chosen were those attuned for the nearly everybody confounders for which penalty were agreed. This advance of deciding on the nearly everyone pertinent consequence permissible the meta-psychiatry to comprise fallout from each learning.

Keywords— Tobacco Smoke, COPD, Data mining, Psychotherapy

1. INTRODUCTION

Sensitivity analyses repeated the complete set of meta-analyses described above for the main index of exposure with the order of preference for time of exposure revised to favor current rather than ever and also preferring unadjusted estimates. For the main exposure index, stepwise regression analysis using forward selection was also used to determine factors independently predicting the risk of heart disease. Similar meta-analyses were also conducted for other indices with sufficient data, though the meta-analyses by subset were more limited. Results of meta-psychiatry are demonstrated in woodland conspire. Inside every scheme, revise guesstimate are planned in mounting regulate of RR. For the foremost catalog, the ballpark figure is an assemblage by setting. Underneath, focusing on a variety of feature of the conclusion and converse probable sources:

1.1 THOSE WHO NEVER SMOKED:

Several revise elucidate that by no means smolder connected to on no account smoking several creation and others that by no means smoking-associated simply to cigarettes. Conversely, many studies merely stated the subjects were never smokers. The distinction is more important in countries where smoking of other products is more common. Some studies also made it clear that the definition allowed the inclusion of those with a limited history of smoking, and a few rejected individuals with cotinine levels typical of current smokers. However, the estimated RR for the main index varied little depending on the definition.

1.2 MISCATALOGING BY NO MEANS SMOKING EMINENCE

No lessons endeavour to conclude whether identity-statement on no account smokers had in actuality smoked in the past. Nevertheless, only a little revise barred those with cotinine echelon analytical of existing smoking. In the modern analysis of lung cancer and ETS, existing investigation representing that modification for miscataloging to a large extent concentrated the anticipated RR for partner's smoking habit. It was not endeavoured such modification here, partially since the degree lies on the extent of the vigorous smoking that is much inferior for heart sickness than for cancer.

1.3 BLUNDER INFORMATIVE ETS DISCLOSURE

Whereas unsystematic mistake in influential ETS disclosure will have a propensity to misjudge any organization with a heart ailment, inaccuracy could not be arbitrary. Hence, lessons of crate-organize or cantankerous-sectional intend, are theme to remember favoritism if the issue with heart sickness is inclined to miscalculate their experience comparative to those devoid of

Devi Rajni, Singh Gurinderpal; International Journal of Advance Research, Ideas and Innovations in Technology heart sickness. Merely two learning has worn biomarker data to endeavor to keep away from recollect preconceived notion. A number of prop up for the continuation of evoke partiality occur from the RRs for the major directory creature advanced for holder-organize and irritated-sectional revision than for potential learn.

2. REVIEW SURVEY

Hutchinson, Natalie (2017) Plasma testosterone and hair cortisol concentrations were measured. In addition, factors related to weight gain after neutering were examined. Increased cotinine concentrations in fur were significantly related to increased percentage weight gain. Several avenues for future research were generated, and many areas warrant further investigation.

Peter N Lee, Barbara A Forey, Jan S Hamling, Alison J Thornton (2017) the tendency for RRs to increase as more factors are adjusted for argues against the association is due to uncontrolled confounding. The increased risk and dose-response for various exposure indices suggest ETS slightly increases heart disease risk. However, heterogeneity, study limitations and possible biases preclude definitive conclusions.

Jesse d. Thacher (2017) Smoke exposure is causally linked to several detrimental health effects and there are no safe levels of maternal smoking during pregnancy or SHS. Despite continued health campaigns, tobacco smoking remains a pervasive problem, and the use of new tobacco products such as e-cigarettes and smokeless tobacco has increased.

Jere Reijula (2015) the risk of cancer among male and female waiters was higher than among the general population in the five Nordic countries. This may be explained by the high prevalence of smoking, heavy occupational exposure to tobacco smoke and high alcohol consumption among the subjects.

Eli Nachamkin (2012) Government efforts to regulate exposure to ETS have occurred at the federal, state and local levels. Additionally, studies should assess whether venues with strict “no-smoking” policies have a greater impact on the smoking behaviors of their employees, both inside and outside of work. Studies should also be conducted to assess the knowledge of venue owners regarding state and local regulations for smoking in public hospitality venues, as not all owners may be aware of the specific requirements.

Noomi Carlsson (2012) with support from CHC nurses, motivated families are able to succeed in their ambitions to make behavior changes in order to protect their children from ETS exposure. The combination of collaborative learning sessions with a “bundle” of evidence-based actions and clinical work has given nurses the experience of being able to lead an improved dialogue with parents and thereby better motivate them to make behavior changes related to smoking.

Peter Boateng Opoku, Heikki Ellilä & Mari Lahti (2011) it would be very commendable if this study area could also be investigated in the third world countries where the impact of passive smoking on the health of children is underestimated due to poor legislation protecting children.

3. PROBLEM DEFINITION

The objectives that are to be focused on this research are:

- To identify the best classifier algorithm into the topics and researches related to environmental exposure.
- Detect the matter of fact that intensifies the idea to get better benefits over the installed system.

4. METHODOLOGY

STEP 1: Data selection.

STEP 2: Data pre-processing.

(a) Splitting training and test datasets.

STEP 3: Feature selection.

STEP 4: Identify the best classification model.

STEP 5: Predicting class levels of test datasets.

STEP 6: Evaluating prediction for credit risk.

5. RESEARCH IMPLEMENTATION

Weka also makes available various data mining techniques like filters, classification and clustering. Here is another example of a data mining technique that is the classification algorithm.

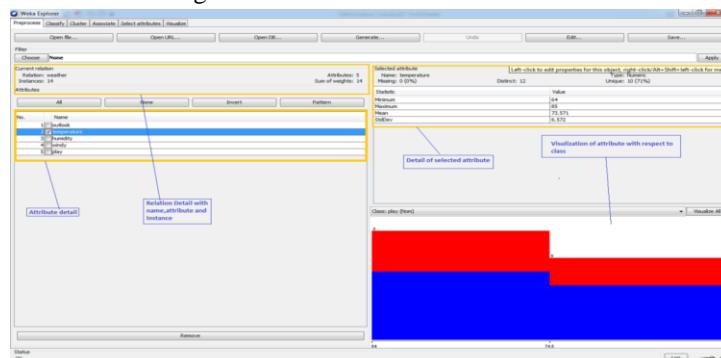


Fig.1: Classification Algorithm

5.1.1 WEKA RESULTS OUTPUT: WRITE PARAMETERS USED

TP = true positives: number of examples predicted positive that are actually positive
 FP = false positives: number of examples predicted positive that are actually negative
 TN = true negatives: number of examples predicted negative that are actually negative
 FN = false negatives: number of examples predicted negative that are actually positive

5.1.2 ACCURACY

Accuracy is how to lock a calculated rate is to the actual rate.
 Accuracy is distinct in the stipulations of correctly confidential occurrence divided by the total number of instances at hand in the dataset.
 Accuracy= (TP+TN)/total
 Where TP- True Positive, TN- True Negative

6. SIGNIFICANT PATTERN AND THEIR CORRESPONDING WEIGHT AGE AND SCORE

Table 1: Significant Pattern and Their Corresponding Weight Age and Score

Parameters	Weight age	Score
Age	<=35	2
	<35-<=75	2
	>75	3
Sex	Male	2
	Female	2
Genetic Risk	No	2
	Yes	1
Obesity	No	2
	Yes	2
Tobacco	Yes	2
	No	3
Smoking	Yes	2
	No	2
Balance Diet	No	2
	Yes	1
Alcohol use	No	1

- Lung Cancer Prediction with High-Risk Level

The screenshot shows a web interface for cancer prediction. It features several input fields with dropdown menus and checkboxes. The fields are arranged in two columns. At the top right, there are three risk level indicators: 'Highest Risk Level (Score: 104)', 'Medium Risk Level (Score: 63)', and 'Low Risk Level (Score: 18)'. The input fields include: Name (Julian Khan), Sex (Male), Age (More than 60), Do You Have Tobacco? (No), Do You Use (Excessive) alcohol? (Yes), Are You Obese? (No), Radiation Therapy in (Chest Area)? (No), Have You Genetic risk? (No), Air pollution? (Yes), Have You balanced diet? (Yes), Occupational Hazard? (No), Are You Smoker? (No), Chronic lung diseases? (No), and Smoker Present Beside You? (Yes). A 'CALC' button is located at the bottom right of the form.

Fig. 2: Experimental result of Cancer Prediction with High-Risk Level.

The experimental results are explained in two parts. One is significant frequent patterns discover and another is shown prediction of Lung Cancer. Using data from the data warehouse, the significant patterns are extracted for Lung cancer prediction. The collected data are pre-processed by deleting duplicate records and adding missing values. Then pre-processed data is clustered using K-means cluster algorithm with k equal to 2. And finally, significant frequent patterns are mined using J48 and Bayesian network.

7. FUTURE SCOPE

Using the two algorithms to determine the best method for predicting lung cancer, we used different attributes such as age(40-70), gender(male/female), genetic risk, cigarettes, tobacco use, alcohol use, diet balance. For those who are smokers and non-smokers, the decision tree algorithm consists of the best-estimated risk of lung cancer. We expected the threat at different levels, including low risk.

8. REFERENCES

- [1] M.S. Jaakkola: Environmental tobacco smoke and health in the elderly.
- [2] European Respiratory Journal 2002 19: 172-181; DOI: 10.1183/09031936.02.00270702
- [3] Environmental tobacco smoke exposure in public places of European cities: Nebot M, López MJ, Gorini G, Neuberger M, Axelsson S, Pilali M, Fonseca C, Abdennbi K, Hackshaw A, Moshammer H, Laurent AM, Salles J, Georgouli M, Fondelli MC, Serrahima E, Centrich F, Hammond SK
- [4] Elizabeth T. H. Fontham, DrPH; Pelayo Correa, MD; Peggy Reynolds, PhD: Environmental Tobacco Smoke and Lung Cancer in Nonsmoking Women: A Multicenter Study
- [5] American Association of Clinical Chemistry. (2012). Nicotine/Cotinine.
- [6] North Georgia Health District (2012). Smoke-free air act.
- [7] United States Department of Health and Human Services (2011). Health consequences of smoking: A report of the surgeon general.
- [8] Allsop, J. & Saks, M. 2007. Researching Health, Quantitative and Mixed Methods, SAGE Publication, Los Angeles, 34-35.
- [9] Abrams, S. M., Mahoney, M. C., Hyland, A., Cummings, K. M., Davis, W., & Song, L., (2006). Early evidence on the effectiveness of clean indoor air legislation in New York State. American Journal of Public Health. 96 (2), pp. 296-298.
- [10] DeLorenze G.N., Kharrazi M., Kaufman F.L., Eskenazi B., and Bernert J.T. Environ Res 2002.
- [11] U.S. Department of Health and Human Services (USDHHS). Women and smoking: a Report of the Surgeon General. USDHHS, CDC, National
- [12] Center for Chronic Disease Prevention and Health Promotion, Office on Smoking and Health: Atlanta, GA, 2001.
- [13] Benowitz N.L. Biomarkers of environmental tobacco smoke exposure. Environ Health Perspect 1999: 107(Suppl 2): 349–355.