

Greenhouse Gas Emission Comparisons with ANOVA

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This report contains an inferential analysis regarding the Greenhouse Gas Emissions from 10 countries between 1990 and 2015. The main analysis aims to find these is a differences of greenhouse gas emissions (kt) across the countries in the past 25 years. ANOVA was performed to determine whether there is any significant difference in greenhouse gas emission among 10 countries, when observed, multiple comperisons on counties were performed. The data is obtained from the Greenhouse Gas Inventory Data of the United Nations Framework Convention on Climate Change.

The 25 years of data shows emissions for each of the nine nations have displayed different patterns, with a majority of nations having very similar emission volumes (Figure 1) The aim of the project is to find out whether there is a significant difference in greenhouse gas emissions among nations. Greenhouse gas emission and countries were analyzed by one-way ANOVA, significance level was set at $P < 0.05$ and pairwise comparisons between the multiple nation's were evaluated.

Our hypothesis are as follows:

H_0 : There is no difference between the amount of Greenhouse Gas Emissions between the 10 nations from 1990 to 2015.

H_A : There is a difference between the amount of Greenhouse Gas Emissions between the 10 nations from 1990 to 2015.

the statistical summaries, and/or the figures in a little report.

term	df	sumsq	meansq	statistic	p.value
Country	9	5.73925E+14	6.37695E+13	4536.75	<.001
Residuals	250	3.51405E+12	1.406E+10	NA	NA

One-way ANOVA indicates that the greenhouse emission of the nine nations are significantly different than each other ($p\text{-value} < 0.05$). Additionally pairwise comparisons were conducted to determine which countries are significantly different from each other (Figure 2? boxplot with the letters)

interprets the findings/results in context of the question.

Latvia demonstrated the least greenhouse gas emmision of all the countries that were included to the analysis, the government of Latvia established policies regarding renewable energy in which was the mair energy consumption by 29.6 % of all energy consumptions in 2009 (Roos et al., 2012). Canada and Japan displayed significantly high greenhouse gas emission in the past 25 years compared to other counties. Jarzen et al. (1998) discussed that 10% of Canada's greenhouse emmision is originated farming-related activities such as farming operations which requires fossil-energy. WHY JAPAN IS HIGH, discuss

Critics, limitations, and assumptions on analysis:

1. European Union data was removed as countries are not identified in the resource and additionally their greenhouse gas emission value was higher that the rest of the countries (Figure 3).

2. If we know the compositions of the EU countries included in this data source, we would be able to split the EU emission into countries and include EU countries into our analysis.
3. In future analysis we should try our best to keep all the original data in our analysis. We could do some sort of transformation to make sense of all data.
4. Future analysis can shed light on greenhouse gas emission over the years and which countries are statistically improving reduction in greenhouse gas emission using time-series analysis.

Appendix

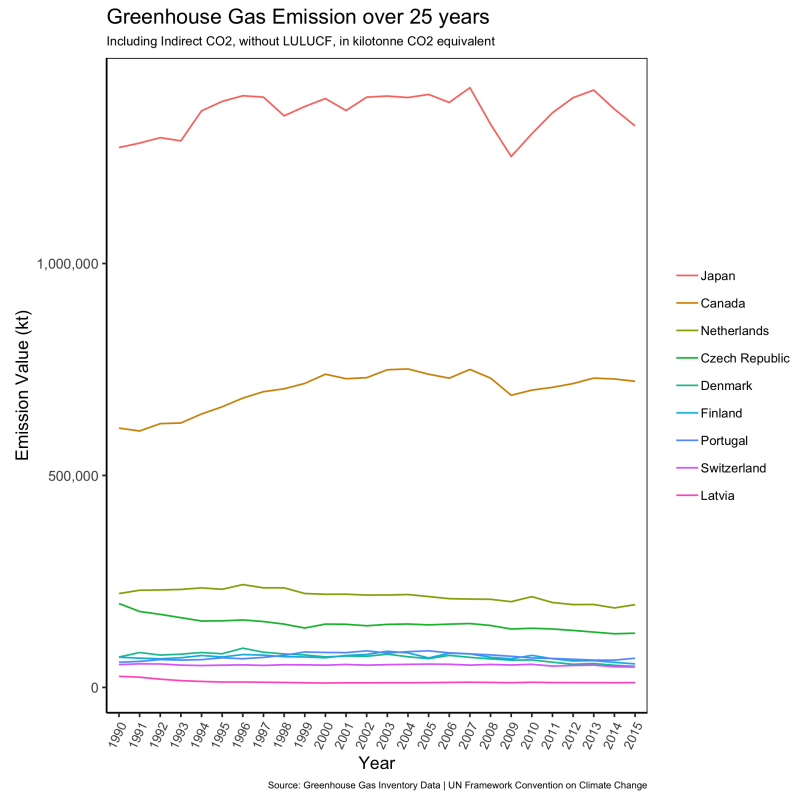


Figure 1: Explainotary data analysis indicates that the European Union had the highest greengouse gas emmision and this category was discarded from the analysis to avoid skewness.

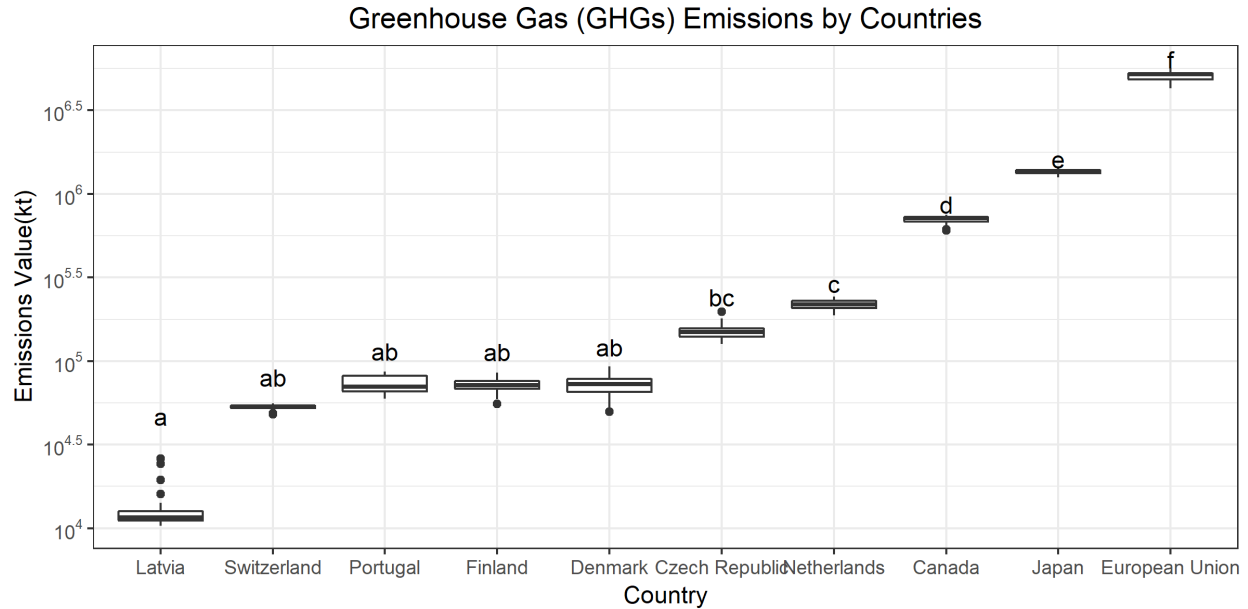


Figure 2: Greenhouse gas emission (kt) of nine different countries in the past 25 years. Different letters indicate significant differences between the groups (pairwise comparison, $p < 0.05$).

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