

Global CO_2 Emissions in 1997

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The year is 1997 and global attention is turning toward the consequences of human-actions in our environmental system. The IPCC has been in existence and studying these trends for more than ten years, and has released its second assessment report in 1995. In this report, the IPCC notes that the balance of the evidence suggests that human-actions play a role in the changing climate. Although, there is little political will to change this activity, neither have global progressive and conservative politicians broken into clear partisan camps. Here, we assess data from the Mona Loa observatory to describe and predict global CO_2 concentrations under several possible scenarios. What we find, when we run the analysis, is going to be grim. Keywords: Replication, Modern Science

I. Introduction

What is the question that you are addressing
why is it worth addressing

What are you going to find at the completion of your analysis.

Understanding a changing climate, and what it means for the earth's inhabitants is of growing interest to the scientific and policy community. Although, at this point in 1997 it is not entirely clear what the consequences of this growing awareness will be, in this report we present likely outcomes under "business-as-usual" scenarios. In doing so, our hope, is to establish a series of possible futures, and, as evidence, technology, and policy develop over the coming decades, that we can weigh the impacts that carbon-emission reduction efforts might take.

A. Carbon Emissions

What are are carbon emissions, and why should anyone care about them? Briefly review what is known about the relationship between the burning of fossil fuels, atmospheric CO_2 What is the current understanding of the linkage between atmospheric CO_2 and global average temperatures.

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II. Atmospheric CO_2 Measurement and Data

Discuss the methods and locations for measurements of CO_2 .

III. Exploratory analysis of historical trends in atmospheric CO_2

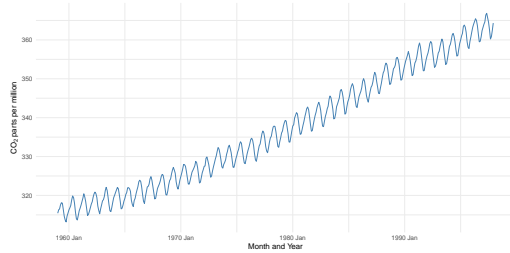


FIGURE 1. HISTORIC TREND IN MONTHLY MEAN $[CO_2]$

Note: CO_2 concentration exhibits accelerating growth, with strong seasonal oscillations. These patterns point to the need of de-seasoning and detrending before further analysis.

Description how, where and why the data is generated

Investigate the trend, seasonal and irregular elements. Trends both in levels and growth rates should be discussed

Atmospheric carbon is plotted in Figure 1, and shows some worrying trends. Just look at how wobbly that line is. How is it possible that we are not living in a simulation, when the lines that plots monthly average CO_2 looks like this?

IV. Models and Forecasts

While these plots might be compelling, it is often challenging to learn the exact nature of a time series process from only these overview, "time vs. outcome" style of plots. In this section, we present evaluate two classes models to assess which time series model is most appropriate to use.

A. Linear and Polynomial Models

Fit a linear time trend model to the 'co2' series, and examine the characteristics of the residuals. Compare this to a quadratic time trend model. Discuss whether a logarithmic transformation of the data would be appropriate. Fit a polynomial time trend model that incorporates seasonal dummy variables, and use this model to generate forecasts to the year 2020.

$$(1) \quad CO_2 = \phi_0 + \phi_1 t + \epsilon_{eit}$$

writing out a model allows you to reference what models your forecasts are being generated from. We will be expecting such a declaration in your reports.

This is a general form of a linear model, where CO_2 concentration is modeled as a linear function of time and a random error.

We estimate best fitting parameters on this model in the following way,

```
model_1 <- lm(y ~ x, data = d)
```

B. ARIMA Models

Following all appropriate steps, choose an ARIMA model to fit to the series. Discuss the characteristics of your model and how you selected between alternative ARIMA specifications. Use your model (or models) to generate forecasts to the year 2022.

C. Forecasts

Because we have fitted a model, we can make predictions from that model. Our preferred model, named in Equation 1 is quite simple, and as you might notice, does not in fact match up with the model that we have fitted. However, from this model it is possible to reason about what the outcomes would be if the *input concept* were to be slightly outside of the observed data range. In particular, if *input concept* were as high as 11, then we would expect the *output concept* to be 10.307, with a prediction interval that ranges from [7.753, 12.86]

V. Conclusions

What to conclude is unclear.

REFERENCES

APPENDIX: MODEL ROBUSTNESS

While the most plausible model that we estimate is reported in the main, "Modeling" section, in this appendix to the article we examine alternative models. Here, our intent is to provide a skeptic that does not accept our assessment of this model as an ARIMA of order (1,2,3) an understanding of model forecasts under alternative scenarios.