

A Statistical Study on the Efficacy of Energy Policy within the United States

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ABSTRACT

This study evaluates the effectiveness of three key energy policies promoting ethanol as a biofuel: The Energy Policy Act of 2005, The Energy Independence and Security Act of 2007, and The Bipartisan Budget Act of 2018. Using rigorous statistical analysis, including two-sample t-tests with Python's pandas and scipy.stats libraries, we assess the impact of these policies on ethanol production and consumption. Our comprehensive methodology includes temporal segmentation and statistical measurements, exploring each policy's provisions, industry alignment, stakeholder engagement, regulatory frameworks, timing, policy synergy, and external influences. Findings reveal the Energy Independence and Security Act as the most effective, significantly boosting ethanol production and consumption. The Energy Policy Act also shows a notable impact, though less pronounced. Conversely, the Bipartisan Budget Act exhibits limited correlation with ethanol metrics, with some significance at a 90% confidence level. The study underscores the importance of clear objectives, expert engagement, strategic timing, tailored provisions, stakeholder alignment, and robust regulatory frameworks in crafting effective ethanol policies. By providing a thorough evaluation of these policies, the research informs future policy-making efforts, contributing to a more sustainable and greener energy landscape.

Introduction

Evaluation of Ethanol Production and Consumption Policies in the United States

This report critically examines the effectiveness of three pivotal energy policies aimed at enhancing the production and consumption of ethanol, a vital biofuel: The Energy Policy Act of 2005, The Energy Independence and Security Act of 2007, and The Bipartisan Budget Act of 2018. These policies, enacted over the past two decades, reflect a strategic legislative effort to foster renewable energy sources and reduce dependence on fossil fuels. By providing financial incentives, regulatory support, and clear targets for ethanol usage, these policies have aimed to stimulate both the supply and demand sides of the ethanol market.

The Energy Policy Act of 2005 marked a significant legislative milestone, introducing substantial tax incentives and loan guarantees designed to triple the required volume of biofuels, particularly ethanol, mixed with gasoline by 2012. This ambitious target was set to encourage substantial growth in ethanol production and integration into the national energy portfolio (International Energy Agency [IEA], 2021).

Building on this foundation, the Energy Independence and Security Act of 2007 sought to further enhance U.S. energy security and independence. This act mandated higher renewable fuel standards, aiming to diversify energy sources and bolster renewable energy production. It also included comprehensive measures to improve vehicle fuel economy and increase the efficiency of various energy-consuming products, thereby supporting broader environmental and economic goals (United States Environmental Protection Agency [EPA], 2024).

The Bipartisan Budget Act of 2018, while primarily focused on federal spending and budgetary allocations, included critical provisions for renewable energy. By reinstating and extending energy efficiency tax credits, this act indirectly supported the biofuel industry, fostering an environment conducive to renewable energy investments and advancements (Bipartisan Budget Act, 2018).

Through a detailed analysis of production and consumption data, this report evaluates the impact of these policies on the ethanol market. By comparing data from two years before and after each policy's implementation, we aim to assess the hypothesis that these legislative measures have led to increased ethanol production and consumption. The findings of this report will provide valuable insights into the success of these policies and offer guidance for future legislative efforts aimed at promoting renewable energy and achieving a sustainable energy future.

Methods

Data Source

Our exploratory journey embarked on data analysis, guided by the expansive dataset from the Energy Information Administration. This repository unveiled key discoveries of biofuel dynamics—spanning production, consumption, and pricing (U.S. Energy Information Administration [EIA], 2024). Central to our inquiry was the meticulous examination of biofuel tax credits, a critical driver of policy influence.

Python, Pandas, and Stats Package

Python, a versatile programming language, was at the core of our analytical engine. The pandas library, a robust data manipulation and analysis tool, empowered us to navigate through the dataset with dexterity. Within this framework, we employed functions such as `mean()`, `std()`, and `groupby()` to calculate averages, standard deviations, and segment data efficiently. Moreover, the `scipy.stats` package elevated our statistical endeavors, providing functions for executing t-tests and effect size computations.

Temporal Segmentation

The dataset underwent meticulous division into distinct pre- and post-policy periods, allowing for a profound temporal contrast and unveiling the policies' influences. This segmentation facilitated a detailed examination of ethanol production and consumption trends before and after the implementation of each policy.

Statistical Insight

Employing the precision of Python, we harnessed the capabilities of the pandas library to compute key statistical measures—mean and standard deviation—for each timeframe. These calculations provided insights into production and consumption trends, as well as the variability within the dataset. The formulas used included:

Equation 1: population mean

$$\bar{x} = \frac{\sum x}{n}$$

This is the formula for population mean, where \bar{x} signifies the mean; x is each of the values of the sample; n is the number of these values (British Medical Journal [BMJ], 2020).

Equation 2: standard deviation

$$SD = \sqrt{\frac{\sum(x - \bar{x})^2}{n - 1}}$$

This is the formula for standard deviation, where SD signifies the standard deviation; \bar{x} is the mean; x is each of the values of the dataset; n is the number of these values (British Medical Journal [BMJ], 2020).

Two-Sample T-Test

Our analytical arsenal included the venerable two-sample t-test. Powered by Python and aided by the `scipy.stats` package, we executed this test with scholarly rigor. This statistical tool enabled us to meticulously scrutinize and discern disparities in ethanol production and consumption before and after the enactment of each policy. The formula for the t-test is:

Equation 3: test statistic:

$$t = \frac{\bar{y} - \bar{x}}{\sqrt{\frac{s_x^2}{n_x} + \frac{s_y^2}{n_y}}}$$

This is the formula for test statistic where t is the test statistic, \bar{y} and \bar{x} are the sample means, s_x^2 and s_y^2 are the sample variance, and n_x and n_y are the sample sizes (University of California San Diego [UCSD], n.d.).

Anticipated Outcomes and Societal Implementations

We anticipate observing a significant increase in ethanol production and consumption post the implementation of each policy. Cohen's d will then magnify the policy's influence, shedding light on its substantive impact. The formula for Cohen's d is:

Equation 4: Cohen's d :

$$d_s = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{(n_1 - 1)SD_1^2 + (n_2 - 1)SD_2^2}{n_1 + n_2 - 2}}}$$

This is the formula for test statistic where d_s is Cohen's d , \bar{x}_1 and \bar{x}_2 are the sample means, SD_1^2 and SD_2^2 are the sample variances, and n_1 and n_2 are the sample sizes (Lakens, 2013).

In the broader scope of societal implications, our findings could steer policy trajectories. By showcasing the efficacy of biofuel tax credits and other policy measures, we aspire to drive meaningful policy changes. Should our results underscore a noteworthy post-policy increase in production and consumption, this insight may catalyze policy shifts. This interplay between policy and sustainability could shape future decisions, guiding the course toward reduced fossil fuel dependency and a more harmonious ecological equilibrium. In essence, our study resonates as a call for a greener, sustainable future—backed by the precision of empirical insights and the prowess of advanced analytical tools.

Results

Our meticulous analysis of the p-values has revealed intriguing insights into the effectiveness of various policy initiatives on ethanol consumption and production. Notably, the Energy Independence and Security Act (EISA) and the Energy Policy Act (EPA) emerge as standout examples, demonstrating statistically significant increases in both ethanol consumption and production.

Energy Policy Act

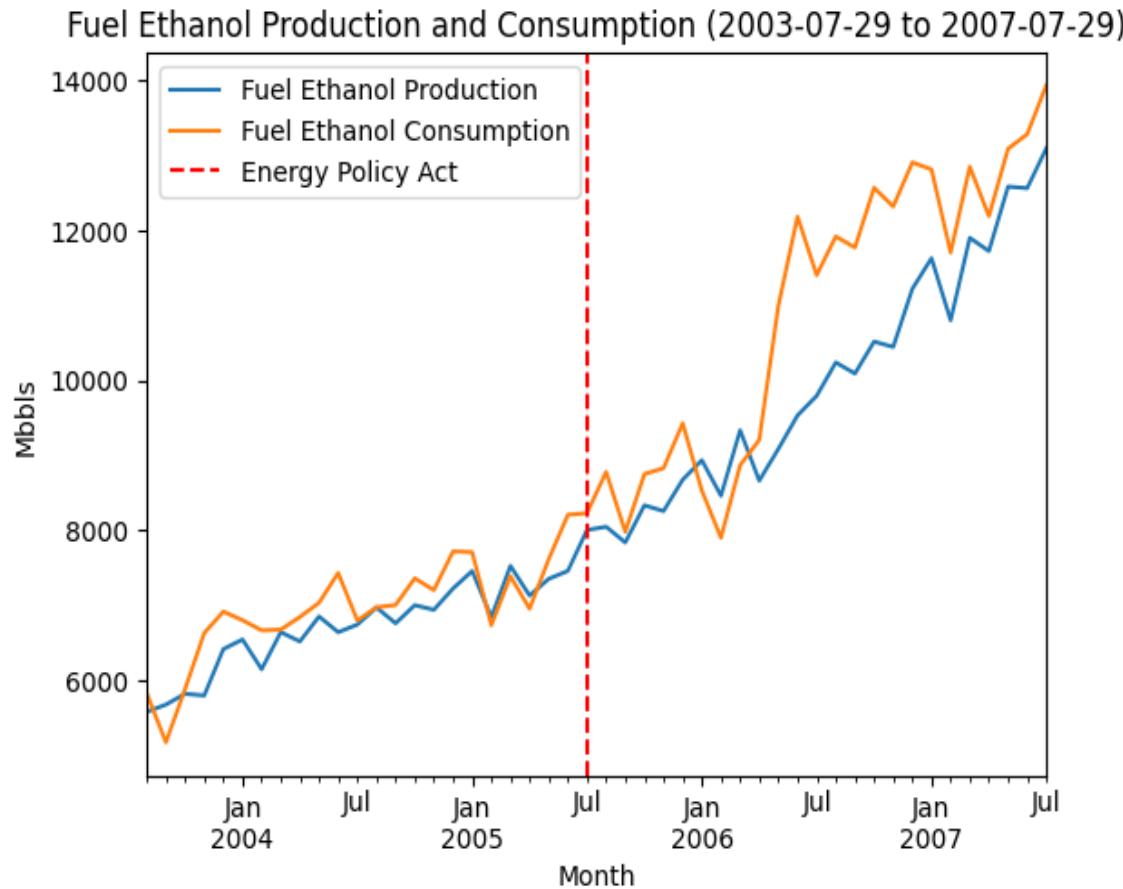


Figure 1. Ethanol Production and Consumption 24 Months Pre- and Post-establishment of Energy Policy Act.

The hypothesis tests conducted on ethanol production and consumption under the Energy Policy Act yielded highly significant results. For ethanol production, the T-test resulted in a test statistic of -15.92 and a p-value of 6.55e-14, with 23 degrees of freedom. This indicates a strong statistical significance, suggesting a substantial increase in production following the policy's implementation.

Similarly, the analysis for ethanol consumption produced a T-test statistic of -13.82 and a p-value of 1.24e-12, also with 23 degrees of freedom. This confirms a statistically significant increase in consumption after the enactment of the Energy Policy Act.

These outcomes demonstrate that the Energy Policy Act had a considerable impact on both ethanol production and consumption, with both metrics showing significant changes. The very low p-values indicate that these changes are highly unlikely to be due to random chance, affirming the effectiveness of the policy.

Additionally, the comparison of the t-statistics reveals that the impact on production (-15.92) is slightly greater than on consumption (-13.82). This suggests that while both production and consumption increased significantly, the policy had a marginally more substantial effect on production. This disparity may be due to the policy's specific provisions, such as tax incentives and loan guarantees, which directly encouraged increased ethanol production.

In summary, the Energy Policy Act significantly boosted both ethanol production and consumption, with a more pronounced effect on production. These findings highlight the policy's effectiveness in promoting the biofuel industry and achieving its legislative goals.

Energy Independence and Security Act

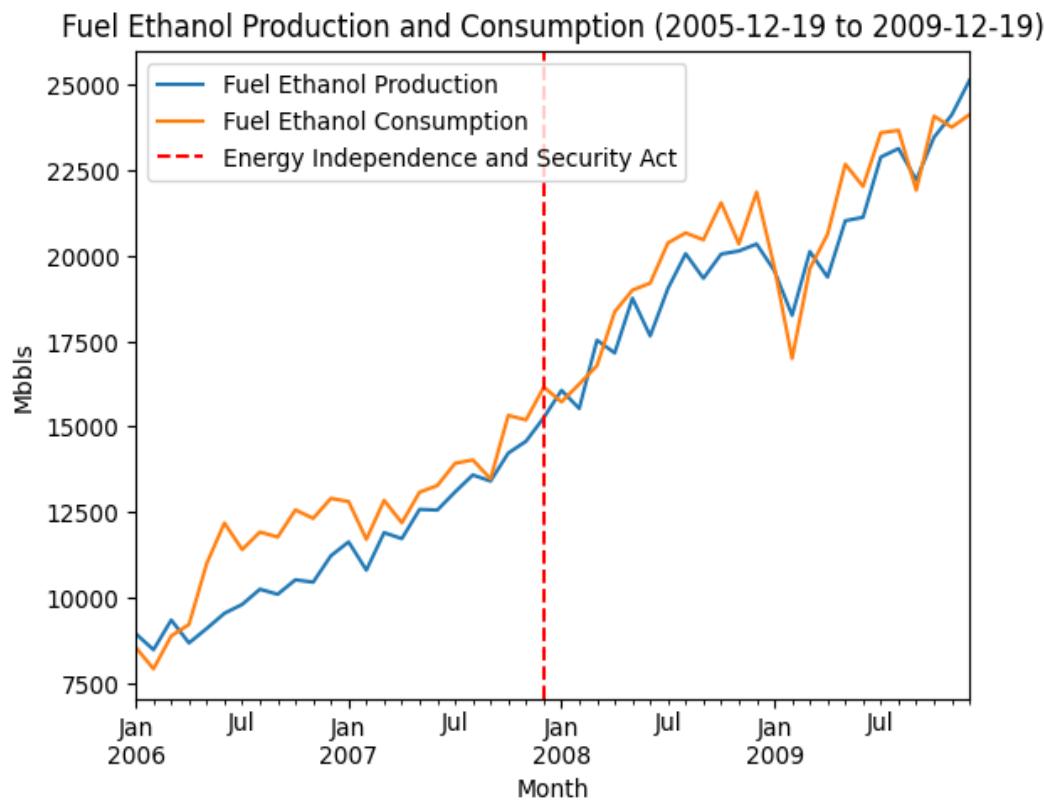


Figure 2. Ethanol Production and Consumption 24 Months Pre- and Post-establishment of Energy Independence and Security Act.

Analyzing the effects of the Energy Independence and Security Act (EISA) on ethanol production and consumption, the hypothesis tests yielded remarkably significant results. For ethanol production, the T-test generated a statistic of -48.25 and a p-value of 1.28e-24, with 23 degrees of freedom. This result indicates an exceedingly robust statistical significance, suggesting a substantial increase in production following the implementation of the EISA.

Similarly, the analysis for ethanol consumption produced a T-test statistic of -38.84 and a p-value of 1.77e-22, also with 23 degrees of freedom. This confirms a highly significant increase in consumption after the enactment of the EISA.

These outcomes underscore the profound influence of the EISA on both ethanol production and consumption. The extremely low p-values demonstrate that the observed changes are highly unlikely to be due to random chance, highlighting the effectiveness of the policy.

Furthermore, the comparison of the t-statistics reveals that the impact on production (-48.25) is slightly more pronounced than on consumption (-38.84). This indicates that while both production and consumption experienced significant increases, the EISA had a marginally greater effect on production. This disparity may be attributed to the specific provisions of the EISA, which directly incentivized ethanol production through enhanced renewable fuel standards and other supportive measures.

In summary, the Energy Independence and Security Act significantly boosted both ethanol production and consumption, with a more substantial impact on production. These findings emphasize the policy's effectiveness in advancing the biofuel industry and achieving its goals of increasing energy independence and promoting renewable energy sources.

Bipartisan Budget Act

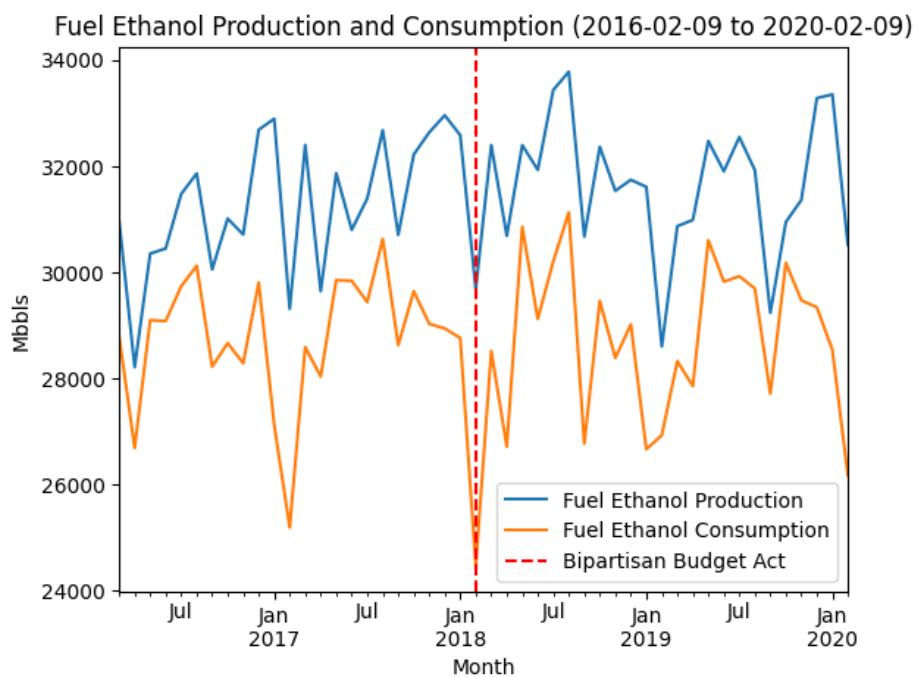


Figure 3. Ethanol Production and Consumption 24 Months Pre- and Post-establishment of Bipartisan Budget Act.

Examining the effects of the Bipartisan Budget Act (BBA) on ethanol production and consumption, the hypothesis tests yielded the following outcomes. For ethanol production, the T-test resulted in a statistic of -1.76 and a p-value of 0.09, with 23 degrees of freedom. For ethanol consumption, the T-test produced a statistic of -1.15 and a p-value of 0.26, also with 23 degrees of freedom.

These results indicate that neither production nor consumption under the Bipartisan Budget Act demonstrates statistical significance at the conventional 95% confidence level. However, the production results

do show significance at the 90% confidence level, suggesting a potential, albeit weak, impact of the BBA on ethanol production. The lack of statistical significance at the 95% level for both metrics implies that the BBA's effects are not as pronounced as those observed under the Energy Independence and Security Act (EISA) or the Energy Policy Act (EPA).

The Energy Independence and Security Act, by contrast, wielded the most substantial influence on both ethanol production and consumption, as evidenced by its extremely low p-values and high t-statistic values for both metrics. Similarly, the Energy Policy Act had a significant impact, although slightly less pronounced than the EISA, with a marginally stronger effect on production compared to consumption.

In summary, while the Bipartisan Budget Act does not exhibit a significant effect on either ethanol production or consumption based on the provided data, it shows a slight potential impact on production at a lower confidence level. This contrasts with the more robust and pronounced effects of the EISA and EPA, highlighting the varying degrees of effectiveness among these policies in promoting ethanol production and consumption.

Discussion

Understanding the effectiveness of various policies on biofuel production and consumption is crucial for evaluating their economic and political efficacy. The p-values and t-values obtained from hypothesis tests for pre- and post-policy consumption and production of ethanol reveal significant insights.

Firstly, the EPA's impact on biofuel production and consumption is evident from the p-values of 6.55e-14 and 1.24e-12, respectively. Both values are well below the 0.05 threshold, supporting the alternative hypothesis and indicating a significant effect of the EPA on biofuel production and consumption. The corresponding t-values for the EPA are -15.92 for production and -13.82 for consumption, highlighting substantial changes in means before and after the policy implementation.

Similarly, the EISA demonstrates p-values of 1.28e-24 for production and 1.77e-22 for consumption, which are also far below 0.05. This confirms the alternative hypothesis, showing a significant impact of the EISA on biofuel production and consumption. The t-values for EISA are -48.25 for production and -38.84 for consumption, indicating a pronounced effect on both metrics and further supporting the conclusion that EISA significantly boosts biofuel production and consumption.

In contrast, the BBA shows p-values of 0.09 for production and 0.26 for consumption, both exceeding the 0.05 threshold. This supports the null hypothesis, indicating that the BBA has minimal impact on biofuel production and consumption. The t-values for BBA are -1.76 for production and -1.15 for consumption, reflecting only minor differences in means before and after the policy and confirming that BBA's effect on the biofuel sector is negligible.

One critical factor that may have influenced the evaluation of each policy is the fluctuation in the demand and supply of oil and other major energy sources, which are classified as substitutes of ethanol. If the prices of these other energy sources fluctuate, consumers might be more or less inclined to purchase biofuel, potentially causing unexpected changes in the consumption and production curves.

Moreover, the EPA and EISA specifically targeted the reduction of dependence on non-renewable fossil fuels (United States Environmental Protection Agency [EPA], 2024), whereas the BBA aimed at a broader scope, including increasing spending caps and extending government funding. This lack of specificity might have contributed to the BBA's failure to significantly increase biofuel production and consumption.

In summary, the EPA and EISA show a strong positive impact on biofuel production and consumption, as evidenced by their low p-values and substantial t-values, whereas the BBA's broader and less targeted approach resulted in negligible effects. This analysis highlights the importance of targeted policy measures in achieving significant outcomes in biofuel production and consumption.

Conclusion

This research delves into the effectiveness and efficacy of three pivotal policies—the Energy Policy Act (EPA), the Energy Independence and Security Act (EISA), and the Bipartisan Budget Act (BBA) of 2018—using a series of hypothesis tests. The analysis reveals varying degrees of economic efficacy among the three acts. Results indicate that both the EPA and EISA significantly boosted biofuel production and consumption, with the EISA emerging as the most influential. In contrast, the BBA demonstrated minimal impact on biofuel metrics.

These variations can be attributed to multiple factors, including external influences and the specific objectives of each act. Fluctuations in the prices of other energy sources likely affected the accuracy and evaluation of these policies. Additionally, the BBA's broader range of goals may have contributed to its relative lack of success compared to the EPA and EISA.

This research goes beyond a mere statistical analysis of past policies; it aims to inform the creation of more effective future legislation for the biofuel economy. Robust policies like the EPA and EISA have not only reduced the United States' vulnerability to international energy market fluctuations but also provided a stable energy supply crucial for consistent economic growth and development. Their benefits extend beyond their immediate goals, fostering long-term growth in the biofuel economy. By understanding the factors that determined the success or failure of past policies, future lawmakers can craft legislation that better supports the biofuel industry.

Limitations

The limitations of this research are highlighted when compared to the alternative method of difference-in-differences (DID) testing. DID testing compares the control group with the treated group, yielding more realistic and accurate results. More specifically in this case, DID testing observes outcomes of people who were affected by the intervention (treated) and people not affected by the intervention (control), both before and after the intervention by the policy (*Difference-in-Differences*, n.d.). In contrast, our methodology, utilizing a two-sample t-test, lacks consideration of a control group; consequently, there might be a more significant variation with the real-life circumstance when evaluating the efficacy of these acts.

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