The Effect of SB1070 on Farm-worker Wages in Arizona

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AZ SB1070 Effect on Wage Replication

This was inspired by Alex Nowrasteh at the Cato Institute. His work on this issue can be found here: https://www.cato.org/blog/wages-did-not-rise-arizona-after-sb1070

Load packages

```
library(stargazer)
library(tidyverse)
library(ggplot2)
theme_set(theme_bw())
```

Read the file into R

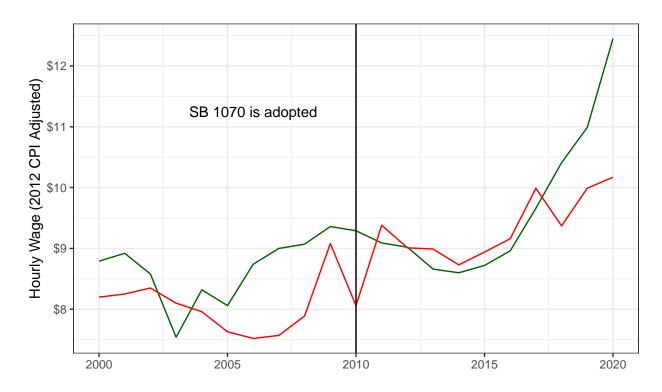
```
fw <- readr::read_csv("/Users/22liddicoat/Downloads/farmwages.csv")</pre>
```

Look at the data

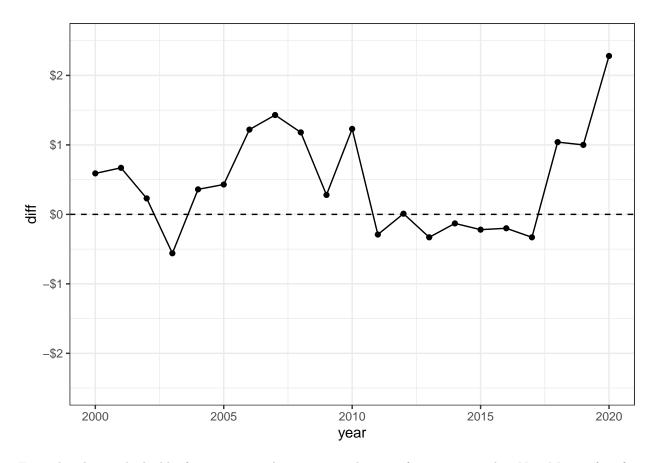
```
fw %>% head()
```

```
## # A tibble: 6 x 4
##
     year state nomwage realwage
    <dbl> <dbl> <dbl>
##
                           <dbl>
                            8.79
## 1 2000 AZ
                   6.56
## 2 2001 AZ
                   6.9
                            8.92
## 3 2002 AZ
                   6.71
                            8.58
## 4 2003 AZ
                   6.02
                            7.54
## 5 2004 AZ
                   6.85
                            8.32
## 6 2005 AZ
                   6.82
                            8.06
```

```
fw %>%
    ggplot(mapping = aes(x = year, y = realwage, color = state)) +
    geom_line() +
    geom_vline(xintercept = 2010) +
    scale_color_manual(values = c("darkgreen", "red")) +
    scale_y_continuous(labels = scales::dollar_format()) +
    labs(x = "", y = "Hourly Wage (2012 CPI Adjusted)") +
    annotate("text", x = 2006, y = 11.25, label = "SB 1070 is adopted") +
    theme(legend.position = "top")
```



```
fw %>%
  select(-nomwage) %>%
  pivot_wider(names_from = "state", values_from = "realwage") %>%
  mutate(diff = AZ - NM) %>%
  ggplot(aes(year, diff)) +
  geom_line() +
  geom_point() +
  geom_hline(yintercept = 0, lty = 2) +
  scale_y_continuous(labels = scales::dollar_format(), limits = c(-2.5, 2.5))
```



From the plots, it looks like farm wages in Arizona started to rise faster compared to New Mexico far after the law was instituted. This suggests that other factors may be at play.

We can play "eye-ball econometrics" all we want, but we can check if our observation is true statistically using a Difference-in-Difference (DiD) regression. More information on this kind of regression technique can be found here: https://www.princeton.edu/otorres/DID101R.pdf

```
did_reg <- lm(realwage ~ time + treated + did, data = df)</pre>
```

```
stargazer(did_reg, title = "DiD Estimates for the Effect of SB1070 on Wages")
```

% Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac at gmail.com % Date and time: Wed, Jul 31, 2024 - 10:04:28

As can be seen in the regression, our coefficient for the DiD variable, which is the variable of interest in our regression, is -0.214 suggesting that wages in Arizona slightly declined relative to wages in New Mexico after 2010. However, since our standard error is 0.480, this gives us a 95% confidence interval of [-1.15, 0.73] $(\beta_{did} \pm 1.96 * SE_{did})$. Since the confidence interval contains 0, we fail to reject the null hypothesis that the law had no effect on wages.

Table 1: DiD Estimates for the Effect of SB1070 on Wages

	$Dependent\ variable.$
	realwage
time	1.199***
	(0.339)
treated	0.583
	(0.347)
did	-0.214
	(0.480)
Constant	8.055***
	(0.245)
Observations	42
\mathcal{E}^2	0.395
Adjusted R^2	0.347
Residual Std. Error	0.776 (df = 38)
F Statistic	$8.264^{***} (df = 3; 38)$
Note:	*p<0.1; **p<0.05; ***p<