Are fire safe cigarettes actually fire safe? Evidence from changes in US state laws

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ABSTRACT
Objective To estimate the effects of fire safe cigarette laws on fire mortality and cigarette-related fires in the USA.
Methods We examined the gradual implementation of the laws to identify their average effects, using difference-in-differences analysis to account for common year effects, time-invariant state effects, state-specific trends and observable time-varying state-level covariates.
Results We found no statistically significant effects on all-cause fire mortality, residential fire mortality or cigarette-caused fire rates. The estimates for cigarette-caused fire deaths were significant under some specifications, but were not robust to the inclusion of state-specific trends or comparisons to effects on other cause-determined fires.
Conclusions Given the mixed state of our results, we conclude that previous claims regarding the effects of fire safe cigarette laws may be premature.

INTRODUCTION
Fires and fire-related mortality is a substantial societal problem across the globe in terms of both destruction of property and loss of life.1 One of the most prominent causes of fatal fires is the ignition of clothing, mattresses or other furniture by smoking materials.2 The population that is affected by these types of events are often considered a marginalised group of individuals that are likely hard to target with conventional prevention strategies (such as information campaigns or smoking bans).3 There is therefore a need for other types of evidence-based prevention measures.

An especially promising universal intervention, at least in theory, is a change in the construction of the cigarette that prevents it from smouldering when not being actively puffed on—a ‘fire safe’ cigarette that passively circumvents the actions of the individuals at risk. The allure of such an intervention is that passively circumvents the actions of the individual, and the behavioural and market responses to the introduction of these laws, that are outside the scope of this study. In summary, they show little to no effects on cigarette sales, intentions to quit or other changes in behavioural responses.18–20 In addition, there appears to be only limited increases in smokers’ exposures to harmful constituents in fire safe cigarettes.16–18

Nonetheless, fire safe cigarette laws are still widely marketed as a successful strategy by, for example, WHO, even though they simultaneously...
request further research on the effectiveness of the intervention. As of 2011, all US states have implemented fire safe cigarette laws. Given the contradictory state of the scientific evidence, we found it relevant to study the case of the introduction of these fire safe cigarette laws using more recent data, and a more comprehensive approach that uses multiple data sets and analytical approaches.

MATERIALS AND METHODS

Data

We used data from several sources to compile a state-years panel data set containing data from 1999 to 2014 for all 50 states and the District of Columbia. Our main outcome measures were all-cause fire mortality, residential fire mortality, cigarette-related fires and cigarette-related fire deaths. For all-cause fire mortality and residential fire mortality, the CDC mortality database provides the most complete records for the study period. International Classification of Diseases codes X00-X09 (exposure to smoke, fire and flames) were used to identify relevant cases (we also used location codes to identify residential fires as this is where most cigarette-related fire deaths occur). However, since the database does not contain information on ignition source, we also turned to the National Fire Incident Data Reporting System (NFIRS) to obtain data on both cigarette-caused fires and fire deaths (as reported by fire departments). All states are covered by the system, but there are two major problems with these data that should be noted. The first is that the coverage is incomplete within each state (fire departments may report to the database voluntarily), and that the number of internally missing values on cause is large (41% for all recorded fires and 54% for fatal fires in the present data set). We have attempted to deal with these issues in our analysis (see below).

Covariate data were also collected from several sources. Data on the policy implementation dates were collected from CDC’s State Tobacco Activities Tracking and Evaluation System, from which we also extracted data on adult smoking prevalence for the period 1998–2011 according to the Tobacco Supplement to the Current Population Survey (TUS-CPS). Since the latter is not conducted on a yearly basis, we also obtained state-years data on per capita taxable cigarette sales (in packs) from Orzechowski and Walker. Per capita ethanol consumption was obtained from the tables in Haughwout et al., per capita disposable income was extracted from the Bureau of Economic Analysis and population data (total population and share of the population aged ≥65 years) was obtained from the U.S. Census Bureau. Summary statistics for all variables are presented in (supplementary appendix table A1).

Statistical analysis

We estimated the effects using Poisson regression models to accommodate for the count nature of all three outcomes (all-cause fire mortality, cigarette-related mortality and cigarette-related fires), using population size as the exposure offset variable. The presence of a fire safe cigarette law per state-year was coded as a dummy variable, where 0 indicates an absence of a state law for state i year t and 1 for time periods when a law is present. To check the sensitivity of the results to model specification, we estimated a range of different models for each outcome, from no trend adjustments, for example, naive before-after analyses, to more complex specifications. We present all these results below to show how the estimates change as we add more variables.

Our preferred specification of these models includes both state and year fixed effects to eliminate the influence of unobserved time-invariant confounders and common, federal time trends (often referred to as a two-way fixed effects model). Under a difference-in-differences framework, these models identify the average causal effect of the laws under a common trends assumption; that is, that there are no state-specific differential trends on the outcome variable. While this may not be true in many real-life applications, there are some potential ways to relax the assumption. One way is to include observable time-varying covariates that could serve as explanations for the diverging trends. To this end, we identified a set of observable explanatory factors that may explain some of the temporal trends based on previous literature. We included per capita disposable income, cigarette sales, ethanol consumption and the per cent of the population aged ≥65 years as observable time-varying covariates. While smoker prevalence might have been preferable to cigarette sales, the prevalence data from TUS-CPS are not available for all state-years during the study period (1999–2014). Instead we used this variable to categorise states into quartiles based on smoker prevalence in order to test for moderating effects of prevalence on the policy impact estimates. For the outcome variables extracted from the NFIRS database, we also included the per cent of events with missing ignition source codes in order to alleviate some of the potential bias stemming from, for example, improvements in cause determination methods over time.

We recognise that it is unlikely that all relevant variables can be observed, and that those we do observe cannot be measured perfectly for the population at risk of cigarette-related fires. We therefore found it necessary to also account for unobservable reasons for differential trends as well. To do this, we included state-specific linear trends to capture some of the differences in trajectory. What the latter option does to the estimate, however, is that it limits its scope to a local effect at the time of the policy change. This can be compared with an abrupt effect in the interrupted time-series framework or a local average treatment effect in the regression discontinuity framework. In essence, it means that for the estimates from a state-specific trends model to be valid, we must assume that the effect occurs abruptly, that is, is not gradual or subject to other changes over time.

A third analysis we employed was to estimate a dynamic effects model, which by inclusion of both leads and lags of the policy variable can serve as a test for whether or not the cause (policy change) occurs before the effect. It can also serve as a validity check for the common trends assumption, where we expect to see a horizontal trend in the effect estimates (with zero mean) the years prior to the policy change. Because of this, we checked for dynamic effects around a span of −4 to +4 years from the intervention since at least 4 years of postintervention data are available for all states. Beyond this time span, the amount of available states that can be used to estimation decreases exponentially since many states have a maximum of 4–5 postintervention years. If periods beyond this are used, biased estimates can be produced if the remaining states differ largely from the mean.

In the cause-specific models, we further complicated the analysis by estimating difference-in-difference-in-differences (DDD) models to compare the effects on affected (cigarette-caused) fires and fire deaths to the contemporaneous effects on cause-determined events that were not related to smoking. Our hypothesis was that if the effects are causal, they should be larger for cigarette-related events than for fires caused by other factors. These models were specified as an extension of the two-way fixed effects model with main effects for state
and year, state-specific trends and time-varying covariates, a law dummy, with the addition of interaction terms between all these variables and a dummy for event type (cigarette-caused or unrelated to smoking).

We present all effect estimates as IRR, which can be interpreted as relative effects. Residual autocorrelation and heteroscedasticity was accounted for using clustered-robust SEs. Uncertainty estimates were derived for the 95% confidence range, assuming a conventional α value of 0.05 as a cut-off for statistical significance.

RESULTS
Effects on all-cause fire mortality
Beginning with the results for all-cause mortality (table 1, panel A), the estimates gave rise to a relatively straightforward interpretation. We found evidence of secular trends in the data, but once these were accounted for, no statistically significant evidence of an effect remained. If anything, there may have been a tendency towards small effects in regions of −5% in all-cause fire mortality. The results did not change noticeably when accounting for the time-varying covariates. Restricting the data to residential fires, where most smoking-related fire deaths occur, did not change the results either (table 1, panel B).

Effects on cigarette-related fires and fire deaths
The results for the effects on cigarette-caused fires show a similar pattern; we found a statistically significant before-after reduction, but it appears that it can be explained, in its entirety, by the presence of secular trends (table 2, panel A, columns I–IV). A further validation of this is the results from the DDD model (table 2, panel A, column V), which shows no evidence of any differences in effect between cigarette-related fires and other cause-determined fires.

Contrary to the other outcomes, the results are more mixed when it comes to cigarette-caused fire fatalities. The results suggest that, while part of the before-after effects can be explained by common year effects and changes in the time-varying covariates, some evidence of an effect remains (table 2, panel B, columns I–III). However, these results are not robust to the inclusion of state-specific trends or compared with reductions in other cause-determined fires (table 2, panel B, columns IV and V).

Dynamic effect estimates
The results from the dynamic effects models for four leads and four lags appear to corroborate the results presented above (figure 1). These models also include year effects, state effects, state-specific trends and per capita disposable income, cigarette sales, ethanol consumption and the per cent of the population aged ≥65 years. The all-cause fire mortality estimates are close to zero for all periods, except for a small positive shock in the period before enactment. The dynamics are similar for residential fire mortality, but the postintervention estimates are slightly larger. Estimates for the occurrence of cigarette-caused fires in general are steadily above the null value in the postintervention period for both the third and fourth lags. Finally, the dynamic effects on cigarette-caused fire fatalities seem to indicate a gradually increasing effect from the time of intervention. However, the deviance from zero in the preintervention periods also gives rise to some concern regarding violations of the common trends assumption. For postintervention estimates to be valid, we would expect the estimates for the lead periods to have zero mean. Instead, it seems they are trending upwards for both cigarette-caused fires and fire fatalities.

The influence of smoker prevalence on the effect estimates
We present effect estimates for state groups categorised by smoker prevalence (in quartiles) to explore if the impact is larger for states with a higher prevalence of smokers in table 3. The results are shown for models with state and year effects only (table 3, panel A) and models that also include state-specific linear trends (table 3, panel B). We obtained similar coefficients from both specifications. The general tendency is the same as in...
the models presented in tables 1 and 2, where we find limited evidence of an effect on all-cause fire mortality, residential fire mortality and cigarette-related fires, but a tendency towards an effect on cigarette-related fire mortality. While almost none of the estimates presented in table 3 is statistically significant, a common feature for all outcomes is that the point estimates are larger in the fourth quartile. It should however be noted that beyond this there is no consistent evidence of a dose–response relationship between smoker prevalence and effect size.

Sensitivity analyses
Estimates from linear fixed effects models gave rise to similar estimates and inferences as in the Poisson regression models presented above. Allowing for more flexible functional forms of the state-specific trends (quadratic and cubic) did not change the results from the models with state-specific linear trends to any meaningful extent. We obtained similar estimates when excluding states with gaps on the outcome variable.

One concern with the results presented thus far is the possibility that late adoption states may already be exposed to the treatment effects through spillovers (in the market share of fire safe cigarettes) stemming from early adoption states. A potential way to test this is to check if adoption order moderates the effect, where we would expect early adoption states to experience a larger impact than late adoption states. To this end, we defined a simple dummy variable for adoption order by dividing the sample in approximately half based on implementation year, where states that enacted the policy before 2009 are defined as early adopters and the rest as late adopters. The results from this exercise showed no indication that the effects were larger in the early adoption states (see online supplementary appendix table A2); in fact, the point estimates are larger for late adoption states for all outcome variables (the groups do not differ significantly from each other on any outcome). This pattern is reversed if we instead define those who adopted fire safe cigarette laws prior to 2010 as early adopters (still, they do not differ significantly).

DISCUSSION
Our results reinforce the mixed state of the evidence regarding the effects of fire safe cigarette laws, and despite finding an indication of an effect on cigarette-related fire mortality rates, we are still left with some unanswered questions. For instance, why would cigarette-related deaths have been affected while the cigarette-related fires in general have remained unchanged? That would imply that even though the quantity of fires is not

![Figure 1](https://example.com/fig1.png)

**Figure 1** Dynamic effect estimates of fire safe cigarette laws in the USA, centred at the year of enactment.

### Table 2 Effects of fire safe cigarette laws on the incidence of cigarette-caused fires

<table>
<thead>
<tr>
<th>Model:</th>
<th>I</th>
<th>II</th>
<th>III</th>
<th>IV</th>
<th>V</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Panel A: cigarette-caused fires</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Effect estimate (IRR, 95% CI)</td>
<td>0.83 (0.71 to 0.96)</td>
<td>1.01 (0.83 to 1.23)</td>
<td>1.02 (0.85 to 1.22)</td>
<td>1.07 (0.87 to 1.31)</td>
<td>1.00 (0.91 to 1.11)</td>
</tr>
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<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
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<tr>
<td>Year effects</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
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<tr>
<td>Time-varying covariates</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>State-specific trends</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>DDD</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>N states</td>
<td>51</td>
<td>51</td>
<td>51</td>
<td>51</td>
<td>51</td>
</tr>
<tr>
<td>Average observations per state</td>
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<td>15.4</td>
<td>15.4</td>
<td>15.4</td>
<td>30.9</td>
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<td>Valid state-years</td>
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<td>787</td>
<td>787</td>
<td>787</td>
<td>1574</td>
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<tr>
<td><strong>Panel B: cigarette-caused fire fatalities</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Effect estimate (IRR, 95% CI)</td>
<td>0.66 (0.55 to 0.79)</td>
<td>0.70 (0.55 to 0.88)</td>
<td>0.74 (0.60 to 0.92)</td>
<td>0.82 (0.66 to 1.02)</td>
<td>0.92 (0.74 to 1.14)</td>
</tr>
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<td>State effects</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
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<tr>
<td>Year effects</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>Time-varying covariates</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
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<td>Yes</td>
</tr>
<tr>
<td>State-specific trends</td>
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</tr>
<tr>
<td>DDD</td>
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<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>N states</td>
<td>51</td>
<td>51</td>
<td>51</td>
<td>51</td>
<td>51</td>
</tr>
<tr>
<td>Average observations per state</td>
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<td>14.9</td>
<td>14.9</td>
<td>14.9</td>
<td>29.8</td>
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<tr>
<td>Valid state-years</td>
<td>759</td>
<td>759</td>
<td>759</td>
<td>759</td>
<td>1518</td>
</tr>
</tbody>
</table>

Poisson regression models with clustered-robust SEs were used to estimate the effects while accounting for correlated errors within states. The dependent variable is offset by population size (per state-year) in all models. The DDD model (column V) includes two observations per state-year; one for cigarette-related events and one for (cause-determined) events unrelated to smoking materials. This model includes main effects for all states, years, time-varying covariates and state-specific trends, as well as their interactions with event type (a dummy for cigarette-caused event or event unrelated to smoking). The time-varying covariates included in models III–V are per capita disposable income, cigarette sales, ethanol consumption and the per cent of the population aged ≥65 years, as well as the proportion of events with unknown ignition source. DDD, difference-in-difference-in-differences model.
affected, the likelihood of dying in a fire is reduced, something that is hard to explain given that the fire safe cigarettes are designed to lower the ignition propensity of cigarettes, which should also affect the likelihood of fires in general. Additional research into the mechanisms behind this difference may therefore be warranted, given that it is not simply a result of bias.

Even if there is some effect of these legislative changes, our study suggests that the magnitude is smaller than previously claimed. For comparison, Alpert et al found that the Massachusetts fire safe cigarette law was associated with a 28% reduction in the risk of cigarette-ignited residential fires (but no significant effects on cigarette-related fire deaths),7 and the 30% decline in smoking-related fire deaths in the USA between 2003 to 2011 has previously been attributed to the concurrent increase in the prevalence of fire safe cigarette laws.8 The European Commission also stated an expected effect of up to −40% in their motivation for enforcing similar laws in the European Union, which is much larger than our estimates suggest.31

**Table 3** The effects of fire safe cigarette laws by smoker prevalence (in quartiles)

<table>
<thead>
<tr>
<th>Smoker prevalence (1999–2011)</th>
<th>Quartile 1</th>
<th>Quartile 2</th>
<th>Quartile 3</th>
<th>Quartile 4</th>
</tr>
</thead>
<tbody>
<tr>
<td>Mean (%)</td>
<td>16.9</td>
<td>19.7</td>
<td>21.3</td>
<td>24.1</td>
</tr>
<tr>
<td>Range (%)</td>
<td>12.4–18.9</td>
<td>19.0–20.5</td>
<td>20.6–21.8</td>
<td>21.9–27.5</td>
</tr>
<tr>
<td>Outcome</td>
<td>Panel A: state and year effects only</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>All-cause fire mortality</td>
<td>0.95 (0.90 to 1.01)</td>
<td>0.95 (0.86 to 1.06)</td>
<td>0.97 (0.84 to 1.12)</td>
<td>0.93 (0.80 to 1.08)</td>
</tr>
<tr>
<td>All-cause fire mortality (residential)</td>
<td>0.98 (0.89 to 1.08)</td>
<td>1.04 (0.80 to 1.34)</td>
<td>0.98 (0.69 to 1.40)</td>
<td>0.91 (0.77 to 1.08)</td>
</tr>
<tr>
<td>Cigarette-related fires</td>
<td>1.02 (0.71 to 1.47)</td>
<td>1.13 (0.98 to 1.30)</td>
<td>1.07 (0.70 to 1.62)</td>
<td>0.86 (0.71 to 1.05)</td>
</tr>
<tr>
<td>Cigarette-related fire mortality</td>
<td>0.68 (0.48 to 0.96)</td>
<td>0.82 (0.52 to 1.27)</td>
<td>0.50 (0.22 to 1.19)</td>
<td>0.55 (0.30 to 1.01)</td>
</tr>
<tr>
<td>Panel B: state effects, year effects and state-specific trends</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>All-cause fire mortality</td>
<td>0.95 (0.90 to 1.01)</td>
<td>0.95 (0.86 to 1.06)</td>
<td>0.97 (0.84 to 1.12)</td>
<td>0.90 (0.79 to 1.03)</td>
</tr>
<tr>
<td>All-cause fire mortality (residential)</td>
<td>0.97 (0.87 to 1.07)</td>
<td>1.01 (0.78 to 1.30)</td>
<td>0.96 (0.70 to 1.31)</td>
<td>0.92 (0.78 to 1.08)</td>
</tr>
<tr>
<td>Cigarette-related fires</td>
<td>1.07 (0.76 to 1.52)</td>
<td>1.08 (0.98 to 1.19)</td>
<td>1.13 (0.82 to 1.54)</td>
<td>0.95 (0.73 to 1.22)</td>
</tr>
<tr>
<td>Cigarette-related fire mortality</td>
<td>0.67 (0.45 to 1.00)</td>
<td>0.95 (0.60 to 1.50)</td>
<td>0.51 (0.15 to 1.68)</td>
<td>0.60 (0.31 to 1.15)</td>
</tr>
</tbody>
</table>

Poisson regression models with clustered-robust SEs were used to estimate the effects while accounting for correlated errors within states. The dependent variable is offset by population size (per state-year) in all models. The coefficients are presented as IRR with 95% CIs in parentheses.

**Limitations**

There are some limitations to our study that should be noted. The lack of in-time randomisation of the legislative actions poses a potential threat to the internal validity of our effect estimates. The models used in the analysis incorporate time-invariant unobservable state and unobservable macro-level confounding factors via the inclusion of state and year fixed effects, as well as within-state changes on observable covariates. As a result, any remaining biases are limited to the state-specific variation on important unobserved confounders, measurement error or other events that are systematically correlated with the enactment of the state laws. However, the results did not change noticeably when state-specific trends were included, which suggests that differential trends may not have been a problem for the purposes of this study. The number of events with missing cause codes in the NFIRS data also calls for some caution in the interpretation of the results related to the cigarette-specific outcomes. While we attempted to alleviate this problem by adjusting for the proportion of events with unknown ignition source, and by comparisons to changes in cause-determined events that were unrelated to smoking in DDD models, these data quality issues may still pose a significant threat to the validity of our estimates. It should also be noted that while our models include state-level covariates, the ecological nature of this data means that we cannot be certain that we have captured the characteristics relevant to the true populations at risk, which is reflected in the fact the estimates do not change once they are included.

**CONCLUSIONS**

We echo the conclusions drawn in a report by the US Consumer Product Safety Commission.14 After finding that fire safe cigarettes appear to be similar to conventional cigarettes in terms of ignition propensity when tested on mattresses, they wrote that ‘... it is premature to conclude that the use of the [fire safe] cigarette alone will greatly reduce the threat of unintentional fires ignited by cigarettes involving mattresses and soft furnishings’. Coupled with our quasi-experimental results, and those of other technical studies,13-17 we find that there is a need for stronger evidence of the effects of fire safe cigarettes on ignition risk in realistic settings before drawing any final conclusions regarding their efficacy.
Correction notice  This paper has been amended since it was published Online First. Owing to a scripting error, some of the publisher names in the references were replaced with ‘BMJ Publishing Group’. This only affected the full text version, not the PDF. We have since corrected these errors and the correct publishers have been inserted into the references.

Contributors  All authors participated equally in the planning and design of the study. NJ designed the empirical strategy along with CB, who collected the data and performed the analyses. All authors contributed to drafts of the paper.

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Competing interests  None declared.

Ethics approval  We used only publicly available, routinely collected (and aggregated) data from administrative registers. Thus, no patients were directly involved in this study, and ethical approval was therefore not required nor sought.

Provenance and peer review  Not commissioned; externally peer reviewed.

Data sharing statement  All data used for this study are publicly available, and the full data set can (and will) be shared upon request from the corresponding author (CB).

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