

1 **Understanding Excess Emissions from Industrial Facilities: Evidence from Texas**

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3 Nikolaos Zirogiannis^{*,+}, Alex J. Hollingsworth⁺, David M. Konisky⁺

4
5 ⁺School of Public and Environmental Affairs, Indiana University Bloomington, Bloomington, Indiana
6 47405, Unites States

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8 ^{*}Corresponding author: 1315 East Tenth Street, Bloomington Indiana-47405, USA,
9 nzirogia@indiana.edu, Tel: 812-856-8323

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Abstract

We analyze excess emissions from industrial facilities in Texas using data from the Texas Commission on Environmental Quality. Emissions are characterized as excess if they are beyond a facility's permitted levels and if they occur during startups, shutdowns, or malfunctions. We provide summary data on both the pollutants most often emitted as excess emissions and the industrial sectors and facilities responsible for those emissions. Excess emissions often represent a substantial share of a facility's routine (or permitted) emissions. We find that while excess emissions events are frequent, the majority of excess emissions are emitted by the largest events. That is, the sum of emissions in the 96th-100th percentile is often several orders of magnitude larger than the remaining excess emissions (i.e. the sum of emissions below the 95th percentile). Thus, the majority of events emit a small amount of pollution relative to the total amount emitted. In addition, a small group of high emitting facilities in the most polluting industrial sectors are responsible for the vast majority of excess emissions. Using an integrated assessment model, we estimate that the health damages in Texas from excess emissions are approximately \$150 million annually.

1 Introduction

Hurricane Harvey made landfall in Texas on Friday August 25th, 2017. Within just a few days, some of the state received over 50 inches of rainfall, inflicting both human casualties as well as devastating infrastructure and property damages. The hurricane caused many industrial facilities in the state to shut down operations and restart them after the rain and flooding subsided. From August 23rd to September 19th, shutdowns and startups (as well as other malfunctions related to Hurricane Harvey) resulted in excess emissions of 1,927 tons of criteria pollutants (616 tons of CO, 735 tons of VOCs, 435 tons of SO₂, 141 tons of NO_x) across the entire state of Texas¹. These excess emissions represent 7% of criteria pollutant excess emissions from all Texas facilities in 2016 due to shutdowns, startups and malfunctions. Although these emissions resulted from a singular, extreme weather event, the same types of emissions (i.e. due to startups, shutdowns, and malfunction) occur on a regular basis during the routine operation of many industrial facilities.

The release of air pollutants is an expected byproduct of many industrial processes. In the United States, the Clean Air Act (CAA) requires that the Environmental Protection Agency (EPA) set ambient air quality standards for a variety of pollutants and emissions limits or technology-based standards for many others. Pollution sources are then required to meet source-specific targets set forth in government-issued permits. Overall, these permits aim to regulate normal operations – for example, the emissions that result from a power plant burning coal to generate electricity or a refinery processing crude oil to make gasoline. However, these types of facilities also often emit “excess emissions,” defined by the EPA as emissions “*that occur during the startup, shutdown, malfunction or other modes of source operation, i.e., emissions that would be considered violations of the applicable emission limitation but for an impermissible automatic or discretionary exemption from such emission limitation.*”².

Excess emissions can be difficult to control because they are attributed to unexpected or unavoidable circumstances. For example, when a facility has to shut down due to a power outage or natural disaster,

55 and then restart its operations, emission levels will likely increase. This increase is due to that fact that
56 pollution control devices require a constant and high temperature environment to function properly³.
57 Similarly, if a control device malfunctions, emissions will increase until the device is repaired. In any
58 case, excess emissions are violations of the CAA per EPA policy because they go beyond authorized
59 limits delineated in a facility's permit².

60
61 In this paper, we use data obtained from the Texas Commission of Environmental Quality (TCEQ) to
62 examine the pattern of excess emissions in the state. We focus on differences across industrial sectors for
63 emissions of criteria pollutants, Volatile Organic Compounds (VOCs) as well as some key hazardous air
64 pollutants (HAPs). In addition, we use topic modeling to delineate the typical causes of excess emissions
65 events, and an integrated assessment model to approximate the monetized value of health damages from
66 the excess emissions that cause either direct or indirect particulate matter (PM).

67
68 Between 2004-2015 excess emissions events in Texas resulted in releases of 104,202 tons of VOCs,
69 which are equivalent to 7.5% of routine (permitted) emissions from all facilities reporting to the state's
70 Emissions Inventory. For CO and SO₂, the relevant figures during the same time-period are 89,202 tons
71 (2.03% of routine emissions) and 123,823 tons (1.84% of routine emissions), respectively. The causes of
72 these events include, unplanned or scheduled startups and shutdowns that exceeded emissions thresholds,
73 weather-induced power outages, and malfunctions due to poor maintenance. We estimate that the
74 monetized value of health impacts, just from direct PM emissions and indirect PM caused by SO₂ and
75 NO_x to be \$148 million in 2015 alone.

76
77 The remainder of the paper is organized as follows. In the next section, we briefly discuss the relevant
78 regulatory history regarding excess emissions, and examine recent changes in EPA policy that affect how
79 the agency interprets excess emission. In section 3, we review the limited existing literature on excess
80 emissions. Section 4 describes the excess emissions data we collected from the TCEQ, and analyzes
81 patterns over time and across sectors, as well as the causes of excess emission events. Last, in section 5,
82 we provide monetary estimates of health damages. We conclude with a discussion of the implications of
83 our research for federal and state policies regarding excess emissions.

84 2 Policy background on excess emissions

85 Air pollution control in the United States is a shared responsibility between the federal government (i.e.,
86 the EPA) and the states. Under the CAA, the EPA sets maximum ambient concentration levels for six
87 criteria air pollutants – CO, lead, NO_x, ozone, PM (2.5 and 10), and SO₂ – as part of the National Ambient
88 Air Quality Standards (NAAQS) program. Based on these standards, the EPA then designates areas of the
89 country meeting the NAAQS as in attainment, and areas failing to meet the NAAQS as nonattainment.
90 Under the CAA, every state is required to submit a State Implementation Plan (SIP) that specifies how it
91 will meet NAAQS requirements, particularly in nonattainment areas. Once the EPA approves a SIP, the
92 state agency begins to implement its strategy, which includes issuing permits to stationary air sources that
93 contain specific emission limits.

94
95 Since the outset of the SIP program in the 1970s, states have incorporated policies regarding excess
96 emissions. Initially the EPA did not consider periods of startup, shutdown and maintenance of equipment

97 as being part of a facility’s “normal” operations. As a result, many SIPs included automatic exemptions
98 and affirmative defense provisions that effectively shielded facilities from enforcement⁴. In a 1977
99 guidance, however, the EPA clarified that automatic exemptions were not allowed. Yet, even though the
100 EPA was concerned that granting sources automatic exemptions for excess emissions would compromise
101 the NAAQS, it did not make enforcement of the 1977 guidance a priority⁵.

102
103 The EPA issued memoranda in 1982 and 1983 clarifying its policies on excess emissions. The agency
104 encouraged states to use an “enforcement discretion approach” to address excess emissions resulting from
105 accidental releases^{5,6}, whereby states could characterize such emissions as allowable if they were clearly
106 due to an unavoidable malfunction. In cases where excess emissions resulted from *scheduled* startup,
107 shutdown, and maintenance activities, the EPA memoranda made clear that these were considered part of
108 the normal operation of a facility, since they were predictable events. Therefore, any excess emissions
109 during scheduled events would be considered a violation of a facility’s permit. This basic approach was
110 reaffirmed and expanded upon in a 1999 memorandum in which the EPA delineated the parameters for a
111 new “affirmative defense” approach that states could incorporate into their SIPs. If specific criteria were
112 met, states could provide air sources with enforcement relief. There were important limitations, however.
113 States could only apply affirmative defense provisions to requests for penalties (not requests for
114 injunctive relief). Moreover, affirmative defense was disallowed for sources that had the potential to
115 cause an exceedance of NAAQS, and to get around any limitations that derive from either New Source
116 Performance Standards or National Emissions Standards for HAPs ^{7,8}.

117

118 2.1 Current policy

119 For about two decades, the EPA did not take significant steps to systematically bring SIPs in line with its
120 own interpretation of excess emissions regulations. In practice, this meant that many states had provisions
121 in their SIPs that conflicted with stated EPA policy, including automatic exemptions and so-called
122 director’s discretion exemptions, which were cases where a state agency head explicitly excused what
123 would otherwise have been a CAA violation. This changed in 2013 when the agency proposed a revision
124 to its policy on excess emissions in response to a petition filed by the Sierra Club. The Sierra Club had
125 argued that many states had provisions in their SIPs that did not observe EPA’s guidance on automatic
126 exemptions, including provisions endorsed by the EPA that limited the jurisdiction of federal courts to
127 impose civil penalties. In its 2015 final ruling on the issue, the EPA found that 36 states had provisions in
128 their SIPs with regards to excess emissions from startup, shutdown and malfunction (SSM) events that
129 were “*substantially inadequate to meet CAA requirements*”². The agency requested that these states revise
130 their SIPs to, where relevant, eliminate automatic exemptions, director’s discretionary exemptions, overly
131 broad enforcement discretion, and all affirmative defense provisions. The elimination of affirmative
132 defense provisions was particularly significant, since the EPA previously allowed them in some
133 circumstances in instances of malfunctions.

134

135 This regulatory history illustrates the complexity and ambiguity surrounding excess emissions. The
136 challenge comes at distinguishing between events that a facility could have reasonably prevented through
137 careful planning and maintenance and those that are truly unavoidable. This complexity is confounded by
138 the lack of data, since with the exception of Texas, Louisiana and Oklahoma, states do not systematically

139 collect data on excess emissions in a fashion that distinguishes them from routine emissions or, if they do,
140 do not make that information publicly available. Consequently, it is virtually impossible to systematically
141 discern the magnitude of excess emissions relative to routine emissions as well as the extent to which
142 excess emissions contribute to poor air quality and adverse health outcomes. In this paper, we explore the
143 most comprehensive dataset on excess emissions, which is based on information reported by major air
144 sources in Texas to the TCEQ. Before turning to these data, we first review the limited existing literature
145 on excess emissions.

146 3 Literature Review

147 Scholars have examined excess emission events in two different ways. First, atmospheric scientists have
148 investigated the degree to which excess emissions of VOCs and NO_x impair air quality. Several studies
149 conducted in the Houston-Galveston region of Texas have found that these emissions, depending on
150 atmospheric conditions, can result in elevated concentrations of ozone⁹⁻¹¹. A separate set of studies have
151 found that excess emissions can result in higher ambient concentrations of fine PM^{12,13}. Furthermore,
152 scholars have recorded large discrepancies between the emissions documented in the Texas emissions
153 inventory and those measured directly through observational studies. The Texas Air Quality Study (Texas
154 AQS) of 2000, for example, found that the state emission inventory underestimated the amount of highly
155 reactive volatile organic compounds (HRVOC) emitted from petrochemical facilities by 1 – 2 orders of
156 magnitude¹⁴. The 2006 Texas AQS found that the emission inventory underestimated emissions for
157 ethene and propene by a factor of 10 and 11 respectively¹⁴. Collectively, these studies provide evidence
158 that excess emissions can have measurable impacts to air quality.

159
160 Most relevant to our work is a second stream of research that has explored patterns of excess emissions
161 across states and industries. A report by the Environmental Integrity Project (2004)¹⁵ was an early attempt
162 to document the way states keep track of and regulate excess emissions. Of the facilities it analyzed, EIP
163 found that CO was the highest emitted pollutant (~48% of total excess emissions) followed by VOCs
164 (~24%), SO₂ (~23%) and very low amounts of NO_x and H₂S. In addition, EIP¹⁵ identified several natural
165 gas plants that in 2003 released excess VOC emissions that were substantially higher than the routine
166 emissions during the previous year. EIP reached similar conclusions in a subsequent report that analyzed
167 data from Texas for the years 2014 and 2015¹⁶. In this study, EIP found that most of the excess emissions
168 of SO₂ and VOCs deriving from malfunctions and maintenance activities were from oil and gas extraction
169 sites, chemical manufacturing plants, oil refineries, and power plants.

170
171 Two other studies have focused their attention on excess emissions from Texas oil refineries. McCoy et
172 al. (2010)¹⁷ find that 96% of reported excess emissions pertained to criteria pollutants (in which they
173 include VOCs), while 63% of the total number of emissions events were concentrated in four areas of
174 Texas (namely Port Arthur, Corpus Christi, Houston and Texas City). The authors also calculated the
175 ratio of upset over total emissions, and found that 30% of the refineries they analyzed had excess SO₂
176 emissions that exceeded 10% of their total emissions.

177
178 Ozymy and Jarrell (2011)¹⁸ conducted a similar analysis for 18 Texas refineries for the 2003-2008 period
179 pointing to the fact that a small number of large excess events released upwards of 500 thousand pounds
180 of SO₂, CO, Propane and Isobutane. When comparing excess and routine emissions from the Toxics

181 Release Inventory (TRI) the authors find that a single excess event can overwhelm the total annual routine
182 emissions of a facility for some toxic pollutants.

183
184 Our paper relies on similar data as much of this prior research, but provides a more comprehensive
185 analysis. Specifically, we analyze excess emissions across sectors, facilities, multiple pollutants, and over
186 more than a decade of time. This approach provides a more complete picture of the nature of excess
187 emissions, in terms of the frequency of occurrence and magnitude of releases. Moreover, we use topic
188 modeling to classify typical common causes of events, as well as an integrated assessment model to
189 provide a monetary estimate of the health damages that can be attributed to excess emissions. We
190 describe the data in the next section.

191 4 Data

192 The data used in this paper come from TCEQ's Emissions Inventory (EI) and Air Emissions and
193 Maintenance Events (AEME) datasets^{1,19}. The former includes annual totals for more than 2,000
194 pollutants from major sources (i.e., CAA Title V facilities) in Texas^{20,21}. Facilities report the following
195 annual amounts in the EI dataset: 1) Routine emissions (i.e., permitted emissions); 2) Emissions events
196 (EE); and 3) Emissions attributed to scheduled startup, shutdown or maintenance (SMSS) events. Taken
197 together, EE and SMSS emissions constitute the total amount of excess emissions. The TCEQ defines an
198 Emissions Event as "*any upset event or unscheduled maintenance, startup or shutdown activity...that*
199 *results in unauthorized emissions*"²². Emissions events result in releases from a stack as opposed to
200 fugitive emissions that "*could not reasonable pass through the stack*"²². An SMSS event, is a scheduled
201 event that is expected to exceed authorized emissions levels and for which a facility is required to provide
202 prior notification and submit a final report to the TCEQ²².

203
204 The TCEQ introduced a rule in 2003 that requires all facilities in the state (not just Title V facilities) to
205 report EE and SMSS emissions within 24 hours of their occurrence provided they surpass an emissions
206 threshold²³. Upon receiving an initial report of an excess emission event from a facility, the TCEQ posts
207 that information on its web-site making it immediately available to the public. The reporting facility has
208 two weeks to submit a final report where it can provide updated information on the event. The
209 compilation of those events across all years (i.e. 2002 until April of 2017) constitute the AEME dataset
210 we obtained from the TCEQ. While facilities are required to report emissions events that exceed a
211 "reportable" quantity in the AEME dataset, the same is not true for the EI dataset. There, Title V facilities
212 are required to report emissions from both "reportable" and "non-reportable" events (i.e. events below the
213 emissions threshold). Because emissions in the EI are reported at the end of the year, facilities might
214 update the excess emissions information they provide to the EI. As a result, at times, there can be
215 discrepancies in the annual sum of excess emissions between the AEME and EI datasets²⁴. Finally, there
216 is no information on routine emissions in the AEME dataset.

217 4.1 Excess Emissions from all facilities

218 Table 1 captures the magnitude and severity of excess emissions in Texas compiling information on EE
219 and SMSS emissions for criteria pollutants, VOCs, and some important HAPs for the period 2004-2015

220 for all Title V facilities. Excess emissions of SO₂ during this period were 123,823 tons, followed by
221 VOCs (104,202 tons) CO (89,202 tons), NO_x (20,227 tons), PM₁₀ (9,572 tons) and PM_{2.5} (6,070 tons).
222 These are large amounts of pollution, accounting for the equivalent of up to 2% of routine (i.e., permitted)
223 emissions for most pollutants, and 7.5% for VOCs. Among the VOCs with highest levels of excess
224 emissions, propane is at the top with 12,081 tons, representing 16.1% of routine emissions. Amongst the
225 463 HAPs, hexane has the highest levels of excess emissions. For all pollutants depicted in Table 1 (with
226 the exception of NO_x) the majority of excess emissions come from Emissions Events (as opposed to
227 SMSS).

228 4.2 Excess emissions by industrial sector

229 As of 2004 there were 3,158 facilities from 231 industrial sectors reporting excess emissions in the EI
230 dataset. Figure 1 displays on the vertical axis tons of excess emissions for criteria pollutants, VOCs and
231 benzene (one of the most prevalent HAPs in excess emissions), while the horizontal axis captures the
232 cumulative number of excess events from all facilities in a given industry. Each circle in the six panels of
233 Figure 1 represents a different industrial sector. The area of each circle captures the ratio of excess over
234 routine emissions for that sector over the 2004-2015 period. The numerical value of the ratio is shown in
235 red for the five sectors with the highest number of cumulative excess emissions across all pollutants. For
236 example, the petroleum refining sector released 18,109 tons of carbon monoxide (CO) excess emissions
237 which represented 10% of the sector's routine emissions during 4,463 events. Of particular interest is the
238 fact that the Natural Gas Liquids sector emitted 77,429 tons of SO₂ during 8,057 events for a ratio of 58%
239 of excess over routine emissions. The refining sector is of particular interest since a very small number of
240 facilities (28 refineries in 2015) release large amounts of excess emissions. The refining sector is the
241 largest emitter of excess VOCs, second largest in excess PM_{2.5}, benzene and SO₂, third largest in CO and
242 fourth in NO_x.

243 4.2.1 Excess emissions from refineries

244 In this section, we take a closer look at excess emissions from oil refineries, an industrial sector that emits
245 disproportionately high levels of excess emissions per facility. Figure 2, provides information similar to
246 that of Figure 1, only this time at the facility level for the top five polluting refineries. The vertical axis in
247 Figure 2 plots routine emissions with each circle being a specific refinery, and the horizontal axis captures
248 the total number of excess emissions events per facility over the 2002-2015 period. The size of each circle
249 captures the ratio of excess over routine emissions. Figure 2 identifies the top five emitting refineries and
250 uses green labels to distinguish them. These five refineries have the highest routine emissions and, often,
251 the largest number of excess emissions events.

252 4.3 Distribution of excess emissions

253 One of the unique characteristics in the pattern of excess emissions is the skewness of their distribution.
254 Figures S14-S17 (in the Supporting Information) plot the skewness parameter of the distribution of each
255 pollutant by industry. In all cases, the values of the skewness parameter are indicative of a distribution
256 that has a small number of events that emit large amounts of pollutants. This fact is further substantiated
257 in Tables S4-S8 (in the Supporting Information) which show the percentiles of the excess emissions

258 distributions by industry. In most industries the median events released less than 1 ton of a pollutant,
259 while the maximum event often released over 1,000 tons. Figures S18-S34 (in the Supporting
260 Information), highlight the fact that a few extreme events dominate the excess emissions distributions.
261 The blue bars in each of those Figures show the total amount of excess emissions (by year) from the
262 bottom of the distribution up to (and including) the 95th percentile. The red bars show the total amount of
263 excess emissions from the 96th percentile to the top of the distribution. The green bars capture the amount
264 released from the single largest event in each year. In the vast majority of cases in Figures S18-S34, the
265 total excess emissions released from the top 5% of the distribution are *larger* than the total excess
266 emissions released from all other events combined. This highlights the fact that several extreme events in
267 each year dominate the distribution of excess emissions. A similar pattern has been documented by
268 Brandt et al. (2016)²⁵ in the case of fugitive methane emissions from natural gas systems, where the
269 largest 5% of leaks represent upwards of 50% of the total amount of leakage.

270 4.4 Important polluters

271 The skewness in the distribution of excess emissions can be traced to events occurring in a small number
272 of facilities. Figure 3, shows total amounts of CO excess emissions for the six most polluting (in terms of
273 excess CO emissions) refineries. The blue dotted line in each panel of Figure 3 traces the annual totals of
274 CO excess emissions in each refinery, while the red solid line shows the number of excess emissions
275 events in each refinery in each year. Those 6 refineries emitted 77% of the total excess CO released from
276 all refineries in Texas between the period 2002-2016. In addition, 35% of all CO excess emissions events
277 that occurred during the same period happened in those six refineries. Figures S35-S38 (in the Supporting
278 Information) illustrate the top six polluting refineries for the remaining criteria pollutants. The Exxon
279 Mobil refineries in Baytown and Beaumont are consistently among the top 6 most polluting refineries in
280 four out of the five pollutants depicted in Figure 3 and Figures S35-S38. Tables S9-S12 (in the
281 Supporting Information), indicate that in all of the top polluting industries, a few key facilities are
282 responsible for the bulk of excess emissions. Detailed information on the top polluting facilities for other
283 key industrial sectors is presented in Figures S39-S50 in the Supporting Information.
284

285 4.5 Causes of excess emissions events

286 An additional piece of information in the AEME dataset is a description of the cause of the excess
287 emissions events, provided by each facility as part of their report to the TCEQ. We analyze these
288 descriptions using a three-step approach that incorporates structural topic modeling (STM)²⁶. First, we
289 find common groupings of words that organize into distinct topics. Second, we determine how well this
290 set of topics explains the observed excess emission descriptions. Finally, we determine the topics that are
291 most likely related to unexpected weather events. The decision to label each topic is validated using a data
292 driven process. Details about the three-step STM approach as well as the validation process are provided
293 in Section 1 of the Supporting Information.
294

295 Figure 4 presents the top five topics by prevalence as well as the prevalence of force majeure weather
296 related topics (e.g., lightning, flash floods, rain, hurricanes, thunderstorms, fires). The most common
297 topics are related to plant shutdowns, flaring, TCEQ reporting terminology, malfunctions, and scheduled
298 maintenance/repairs. On average, around 5% of event description text is related to shutdowns. Text
299 related to weather induced accidents composes just over 10% of all comment text, indicating that while

300 weather related accidents are an important source of emissions events, they are far from the dominant
301 source.

302

303 5 Monetary estimates of health damages from excess emissions

304

305 While the adverse health effects of excess emissions are likely significant, their precise empirical
306 estimation would require access to detailed data on mortality, morbidity, and pollution exposure. We use
307 an integrated assessment model (IAM) to calculate a “back of the envelope” estimate of the monetary
308 value of the health effects from excess emissions. To conduct this analysis, we first aggregate all the
309 annual emissions of PM_{2.5}, SO₂, and NO_x from the EI dataset to the county level for the period 2004-
310 2015. These annual county level emissions are put into the Estimating Air Pollution Social Impact Using
311 Regression (EASIUR)^{27,28} model, which predicts the total damage from a marginal increase in pollution
312 from any county in the continental U.S. The damage estimates provided by the EASIUR model are from
313 the perspective of the source county, where total damage from each source county’s pollution is the
314 aggregation of damage done by that source county on all receptor counties. The EASIUR model is based
315 on the Comprehensive Air Quality Model with Extensions (CAMx) and it’s damage predictions compare
316 well with the results from other IAMs. The marginal damage estimate of the EASIUR model is based
317 upon the impact of directly and indirectly emitted PM_{2.5} on mortality. Estimates include damages that
318 occur both locally and in downwind regions. In addition to varying across geographic space, predicted
319 marginal damages vary with seasonal patterns in pollution transport, stack emission height, and pollutant
320 type (PM_{2.5}, SO₂, NO_x, and NH₃).

321

322 We find that in 2015 excess emissions are responsible for at least \$148 million in health damage annually,
323 with approximately 10% of this damage coming from oil refineries. The EASIUR model uses a value of
324 statistical life (VSL) estimate of \$8.8 million implying that 16.82 deaths are caused per year by excess
325 emissions. Damages vary across the state, and are concentrated in areas with more large, industrial
326 facilities. Figure 5 displays the county-level damages estimated by EASIUR for 2015. Figures S52-S55 in
327 the Supporting Information provide county level damages by pollutant by year. Figure S56 in the
328 Supporting Information shows aggregate annual health damages by year from all sources and from the
329 refining sector specifically. It is important to emphasize that our damage estimates represent only those
330 mortality impacts due to direct and indirect PM_{2.5} emissions. Thus, they are intended to serve as a
331 conservative, lower bound for potential health damages. Excluded health damages include all acute health
332 events that do not lead to mortality and all pollution induced mortality that is not related to PM_{2.5}.

333

334 6 Conclusion

335

336 This study examines the significance of excess emissions, a category of air pollution that has received
337 little attention in the scholarly literature. Our analysis shows that excess emissions are not exceptional,
338 outlier events, but rather a regular feature of operations at industrial facilities. The data reported to the
339 TCEQ show that these emissions can also be substantial in magnitude, raising important questions for
340 future research about their effects on air quality and public health. In addition, in the most polluting
341 industries, a small group of facilities are responsible for the vast majority of criteria pollutant excess

342 emissions. Given the significant public health impacts, which we estimate to be at least \$148 million
343 annually in Texas alone, it is imperative that all states begin to systematically track excess emissions.
344

345 Excess emissions are also important from a policy perspective. As discussed above, the EPA has recently
346 revised its policy on how excess emissions are regulated under the CAA. The agency has always regarded
347 these emissions as a violation of a facility's permit obligations under the statute². However, enforcement
348 has largely been left to the states, and the EPA has determined that too often states have relied on policies
349 and procedures that inappropriately shield firms from penalties. The EPA is in the process of reviewing
350 many states' SIPs to ensure that treatment of excess emissions is consistent with EPA's interpretation. As
351 is the case with many EPA policies, the Trump Administration is now reviewing the policy itself, which
352 leaves the question of how excess emissions will be handled under the CAA in the future. Given the
353 importance of these emissions, these policy decisions will be consequential.
354

355 **Supporting Information**

356 The Supporting Information is available free of charge on the ACS Publications website:

357 Additional tables and Figures

358 Details on the three step Structural Topic Modeling Approach
359

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Tables and Figures

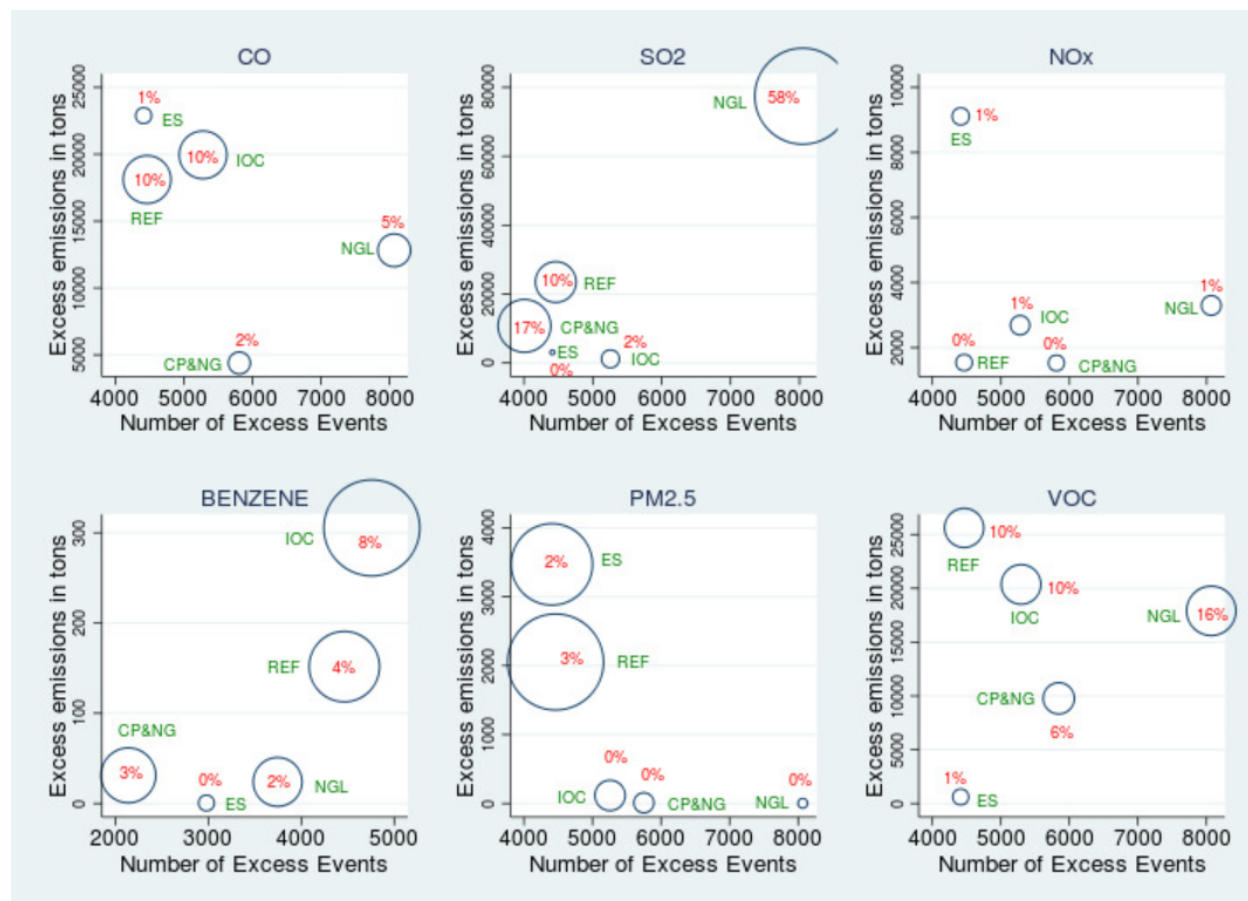
Table 1: Total amounts of excess emissions (in tons) from all facilities reporting to the Emissions Inventory of the TCEQ during the period 2004-2015. Source: Table compiled by the authors using data from TCEQ¹⁹.

	Contaminant	Total excess emissions (tons)	Total excess / Total Routine	Emissions Events / Total excess emissions
Criteria Pollutants	Sulfur Dioxide	123,823	1.84%	82.91%
	Volatile Organic Compounds	104,202	7.50%	72.74%
	Carbon Monoxide	89,202	2.03%	57.18%
	Nitrogen Oxides	20,277	0.47%	43.16%
	Particulate Matter 10	9,572	1.37%	54.71%
	Particulate Matter 2.5	6,070	1.46%	53.77%
VOCs	Propane	12,081	16.09%	84.56%
	Propylene	6,527	19.70%	83.38%
	Isobutane	4,632	13.43%	82.47%
	Butene	734	16.26%	83.20%
HAPs	Hexane	2,150	8.42%	54.33%
	Toluene	840	4.53%	58.65%
	Benzene	776	5.53%	60.23%
	Xylene	318	2.06%	56.02%
	Formaldehyde	70	0.22%	66.40%

Note: Between 1990 and 2003, facilities reported: a) total excess emissions, that is, Emissions Events (EE) + emissions from Scheduled Maintenance Startup and Shutdown (SMSS) and b) routine emissions. Starting in 2004, amounts were reported separately for each of the three categories of emissions (i.e. routine, EE and SMSS emissions). We do not report data on lead because of low levels of emissions, and ozone which is not directly emitted. The column labeled “Total excess emissions (tons)” shows the total amount of EE+SMSS. The column labeled “Total excess over Total Routine” shows the ratio of Total excess emissions (EE+SMSS) over routine emissions. Finally, the column labeled “Emissions Events/Total Excess emissions” shows the ratio of emissions events (EE) over total excess emissions (EE+SMSS).

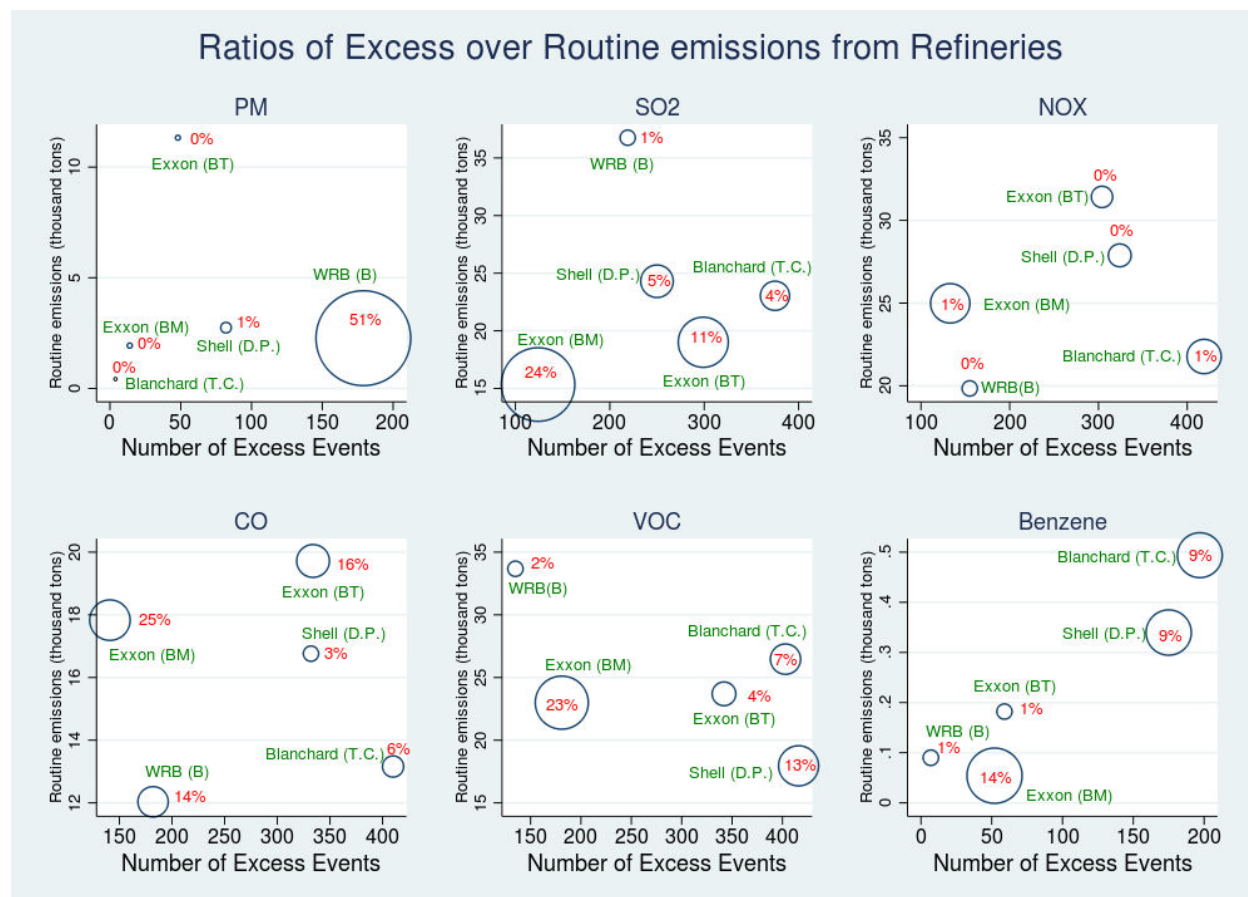
A more detailed version of this table is provided in Table S1 of the Supporting Information. Figures S1 and S2 of the Supporting Information, provide information on ratios of excess over routine emissions by year for criteria pollutants and important HAPs. Tables S2 and S3 of the Supporting Information provide data on excess emissions by industrial sector, while Figures S3-S8 provide time trends of excess emissions of criteria pollutants by industrial sector.

Figure 1: Ratio of excess emissions over routine (permitted) emissions (captured in red), total amount of excess emissions (on the vertical axis), and total number of excess emissions events (on the horizontal axis) from the top 5 polluting industries during the period 2004-2015. Source: Figure compiled by the authors using data from TCEQ¹⁹.



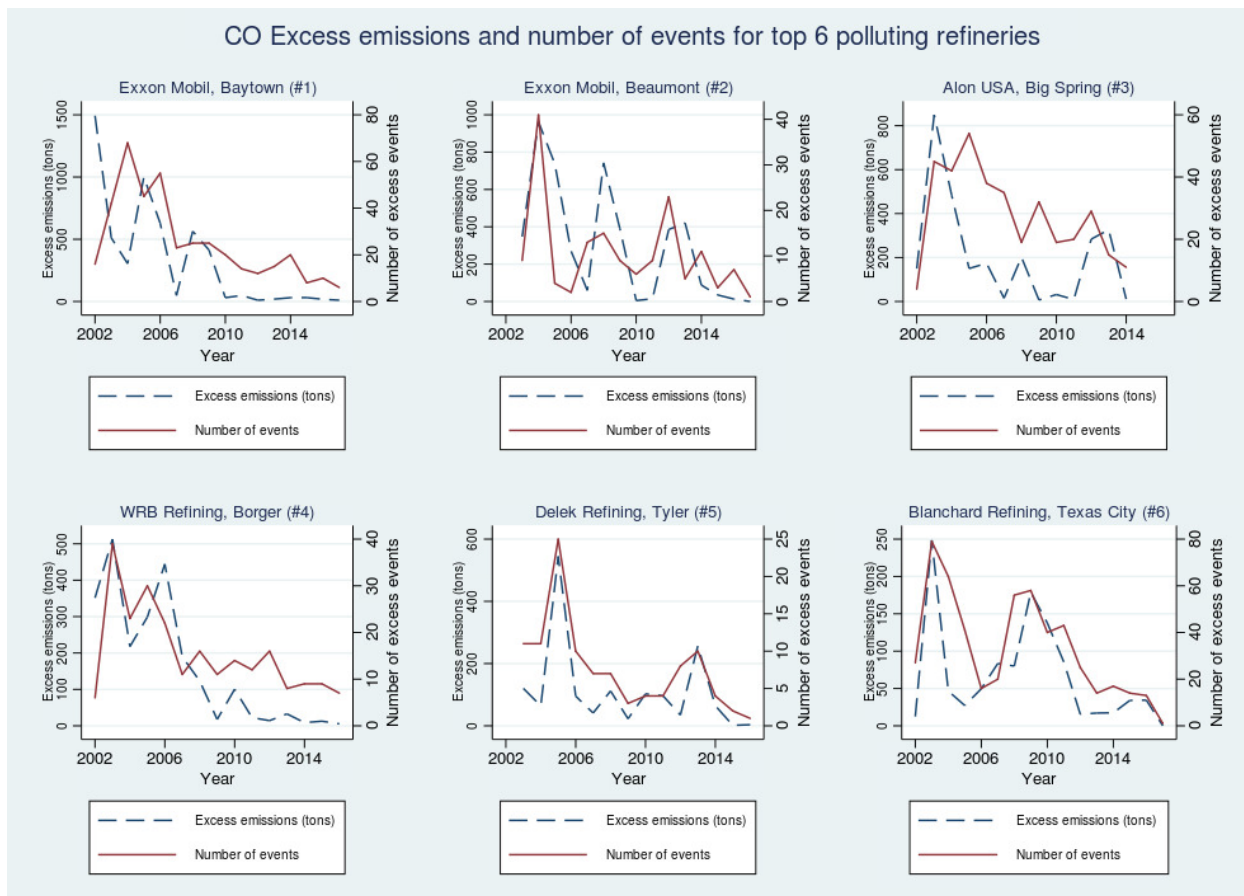
Note: The five industries with the highest amount of cumulative excess emissions across all pollutants (as identified in Table S2 of the Supporting Information) are highlighted in green. Those industries are: a) Crude Petroleum and Natural Gas (CP&NG), b) Natural Gas Liquids (NGL), c) Petroleum Refining (REF), d) Industrial Organic Chemicals (IOC), e) Electric Services (ES). The ratio of excess over routine emissions is captured in red and is also depicted by the area of each circle (the larger the circle, the higher the ratio of excess over routine emissions). Note that the areas of each circle are not comparable across the six panels of Figure 1, but are comparable within each of the six panels. The number of excess emissions events in the horizontal axis comes from the EI dataset and includes both reportable and non-reportable events by facility between 2006-2015 (there is no information on counts of events for prior years in the EI dataset).

Figure 2: Ratio of excess emissions over routine (permitted) emissions (captured in red), total amount of routine emissions (vertical axis), and total number of excess emissions events (horizontal axis) from the top 5 polluting refineries in Texas, during the period 2002-2015. Source: Figure compiled by the authors using data from TCEQ^{1,19}.



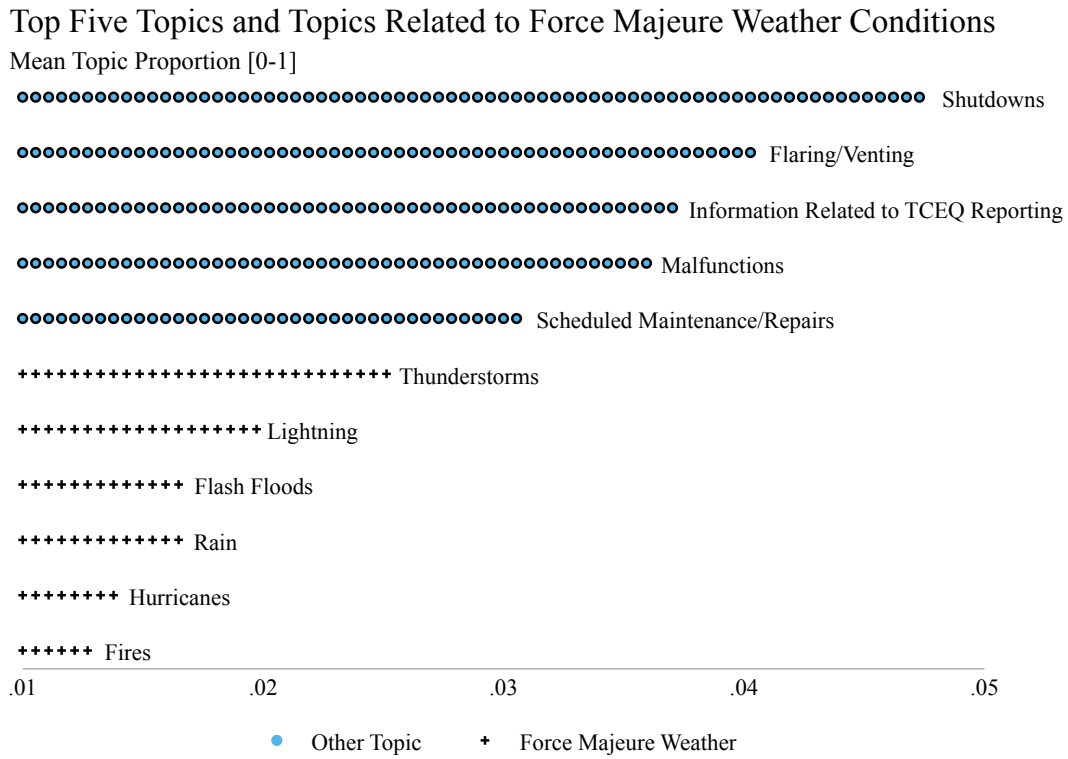
Note: The refineries with the highest amounts of cumulative excess emissions are highlighted in green. These are: a) Shell Oil, Deer Park (D.P.), b) Exxon Mobil, Baytown (BT), c) Blanchard, Texas City (T.C.), d) WRB, Borger (B), e) Exxon Mobil, Beaumont (BM). The ratio of excess over routine emissions is captured in red and is also depicted by the area of each circle (the larger the circle, the higher the ratio of excess over routine emissions). Note that the areas of each circle are not comparable across the six panels of Figure 2, but are comparable within each of the six panels. Figure S10 in the Supporting Information is a version of Figure 2 that includes all Texas refineries. Figures S11-S13 in the Supporting Information provide similar information for other key industrial sectors (namely, Crude Petroleum and Natural Gas, Industrial Organic Chemicals, and Natural Gas Liquids). Contrary to Figure 1, the number of excess emissions events in Figure 2 comes from the AEME dataset and covers the period 2002-2015.

Figure 3: Excess emissions and number of excess emission events involving the release of Carbon Monoxide (CO) for the top 6 most polluting refineries. In each of the six panels, the amount of excess emissions for each facility is measured on the left axis and illustrated with the blue dotted line, while the number of excess emission events for each facility is measured on the right axis and illustrated with the red solid line. The name and location (city) of each facility are listed in the title of each panel. Those 6 facilities released 77% of all CO excess emissions across all refineries between 2002-2016. The total number of excess emissions events from those 6 refineries represent 35% of all excess emissions events from all refineries between 2002-2016. Source: Figure compiled by the authors using data from TCEQ¹.



Note: Figures S35-S38 in the Supporting Information present similar information as that depicted in Figure 3 for the remaining criteria pollutants on the top 6 polluting refineries. Figures S39-S50 in the Supporting Information have data on facilities in the most polluting sectors (namely, Crude Petroleum and Natural Gas, Natural Gas Liquids, and Industrial Organic Chemicals). Tables S9-S12 in the Supporting Information show summary statistics for the top six polluting facilities in each of the top four polluting sectors.

Figure 4: Results of Structural Topic Modeling. Source: Figure compiled by the authors using data from TCEQ¹



Note: The prevalence and full set of all 50 topics estimated is available upon request.

Figure 5: Damages from Excess Emissions in 2015, by Texas County. Source: Figure compiled by the authors using data from TCEQ¹⁹, EASIUR²⁷, QGIS²⁹ and Manson et al (2017)³⁰

