

Tradeoffs Among Indoor Air Quality, Financial Costs, and CO₂ Emissions for HVAC Operation Strategies to Mitigate Indoor Virus in U.S. Office Buildings

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Abstract

Adapting building operation during the COVID-19 pandemic to improve indoor air quality (IAQ) while ensuring sustainable solutions in terms of costs and CO₂ emissions is challenging and limited in literature. Our previous study investigated different HVAC operation strategies, including increased filtration using MERV 10, MERV 13, or HEPA filters, as well as supplying 100% outdoor air into buildings for a system initially sized for MERV 10 filtration. This paper significantly extends that research by systematically analyzing the potential financial and environmental impact for different locations in the US. The previous medium office building system model is improved

to account for operation in different climates. New evaluation metrics are created to consider the comprehensive impact of improving IAQ on costs and CO₂ emissions, using dynamic emission factors for electricity generation depending on the location. HVAC operation strategies are studied in five different locations across the United States, with distinct climates and electricity sources. In four of the five locations, MERV 13 filtration offers the best improvement in IAQ per increase in costs and emissions relative to MERV 10. The exception is the mildest climate of San Diego, where use of 100% outdoor air provides the best IAQ with a limited increase in costs and emissions. A system not sized for HEPA filtration can lead to increased costs and emissions without much improvement in IAQ.

Keywords: Indoor air quality, financial cost, CO₂ emissions, COVID-19 pandemic, climate change.

1. Introduction

Sustainably operating buildings to improve indoor air quality (IAQ) is critical during both a global pandemic and rapid climate change. The United States (U.S.) is the second highest contributor to global greenhouse gas emissions [1] and buildings account for about 36% of energy-related CO₂ emissions in the U.S. [2]. Building operation during the COVID-19 pandemic

7 is crucial, as studies have shown that the risk of infection indoors caused by
8 airborne transmission is significant [3, 4, 5]. Strategically operating building
9 heating, ventilation, and air-conditioning (HVAC) systems can improve IAQ
10 and reduce the risk of infection from airborne viral particles [6, 7, 8, 9],
11 but can also result in increased energy consumption [10, 11]. This can be
12 caused by increased fan energy to overcome the additional pressure drop
13 of more efficient filters, or increased heating and/or cooling energy due to
14 higher outdoor air ventilation rates, for example. Balancing both IAQ and
15 sustainability is a challenge that depends on many factors such as mitigation
16 strategy, climate, energy sources, etc.

17 Previous research has attempted to study the tradeoffs between IAQ
18 and sustainability for various mitigation strategies and climates. Schibuola
19 and Tambani [12] studied using increased mechanical ventilation with high
20 efficiency air handling units to reduce the risk of infection of COVID-19
21 and improve energy efficiency in Italian secondary schools. They found
22 increasing mechanical ventilation can significantly reduce infection risks, and
23 the increased energy can be offset via the installation of high efficiency air
24 handling units. Sha et al. [13] investigated increasing building ventilation
25 while reducing energy consumption via direct cooling with outdoor air in

26 high rise buildings, and found that improving the ventilation control allowed
27 for around 40% reduction in energy consumption while meeting required
28 ventilation rates. Zaatari et al. conducted multiple studies [14, 15] investigating
29 the tradeoffs of IAQ and energy consumption for different levels of filtration
30 or ventilation. They found that the best filtration or control strategy is
31 dependent on building system and climate. Santos and Leal [16] studied the
32 impact of ventilation rate on energy consumption in European climates and
33 found increasing ventilation rate can significantly increase energy consumption.
34 Ben-David and Waring [17] compared the associated costs for different levels
35 of filtration and ventilation for office buildings in different climates. The
36 results showed that improving filtration and increasing ventilation rate complement
37 each other, and improving filtration tended to have a greater impact on the
38 cost function. Our previous work [10] created new component models for
39 HVAC filters and viral transmission and implemented them in a dynamic
40 system model using Modelica language. The new models were used to analyze
41 indoor virus concentration, predicted number of infections, and energy consumption
42 for different mitigation strategies, including use of 100% outdoor air and
43 MERV 10, MERV 13, and HEPA filtration.

44 Although significant progress has been made, further analysis can be

45 performed to understand the tradeoffs between IAQ and sustainability as the
46 pandemic enters its third year. First, some studies may investigate HVAC
47 operation strategies in different climates, but do not always consider the
48 differences in operation based on climate. For example, buildings in humid
49 climates operate their systems differently in unoccupied hours to avoid build
50 up of mold. Furthermore, studies often assume constant outdoor airflow
51 rates and do not account for dynamic outdoor airflow rates based on the
52 control of the airside economizer. The amount of free cooling provided by
53 the airside economizer impacts both IAQ and energy consumption and varies
54 among climates. Also, studies often quantify sustainability in terms of energy
55 consumption or cost, but greenhouse gas emissions are not always considered.
56 This becomes especially important in the age of rapid climate change, since
57 building operators may prioritize minimizing greenhouse gas emissions over
58 IAQ or energy costs. New policies may also incentivize limiting greenhouse
59 gas emissions by placing a tax on these emissions. Furthermore, new metrics
60 are needed to quantify the tradeoffs among IAQ, costs, and emissions.

61 To address this research gap, we propose a study to analyze the tradeoffs
62 among IAQ, financial costs, and CO₂ emissions of four mitigation strategies in
63 five unique geographic locations with distinct climates and electricity sources

64 across the U.S. Five of the 17 sustainable development goals outlined by the
65 United Nations [18] are targeted in this paper: 3) good health and well-being,
66 7) affordable and clean energy, 9) industry, innovation, and infrastructure,
67 11) sustainable cities and communities, and 13) climate action. The studied
68 mitigation strategies include different levels of filtration, such as MERV 10,
69 MERV 13, and HEPA filtration, as well as supplying 100% outdoor air
70 with MERV 10 filtration. We simulate the scenarios using detailed system
71 modeling of a prototype medium office building initially sized for MERV 10
72 filtration based on the Modelica *Buildings* library [19, 20]. Our scientific
73 contributions in this paper include: 1) developed detailed system models
74 to account for the dynamics of the HVAC system to simulate mitigation
75 strategies in different locations with distinct climates, 2) proposed novel
76 comprehensive evaluation metrics which consider the effectiveness of mitigation
77 strategies in terms of IAQ, financial costs, and CO₂ emissions, including using
78 newly available dynamic CO₂ emission factors dependent on location, and 3)
79 identified mitigation strategies in each location that improve IAQ by 6-16%
80 with limited increases in costs and emissions.

81 The remainder of this paper is organized as follows. We introduce the
82 building system model and improvement to account for operation in different

83 climates in Section 2. Next, methods to evaluate and compare the mitigation
84 strategies are detailed in Section 3. The scope of analysis for this study
85 including the four mitigation strategies and five locations is described in
86 Section 4. The results in terms of IAQ, costs, and CO₂ emissions are presented
87 in Section 5. Finally, conclusions are drawn in Section 6.

88 **2. Building System Modeling**

89 We first introduce the medium office building system studied in this
90 paper. The system modeling for the different climates is then detailed.

91 *2.1. Building System*

92 The studied building is based on the DOE commercial reference medium
93 office building [21], with a focus on the bottom floor based on an existing
94 model [22]. The schematic for this system is shown in Figure 1. The floor
95 consists of five zones, including a core zone and four perimeter zones. A
96 central air handling unit with heating and cooling coils services this floor,
97 with VAV terminal boxes containing reheat coils for each zone. An outdoor
98 air economizer is used to supply the minimum outdoor airflow based on
99 ASHRAE standards [23] as well as provide free cooling. Natural gas is used to
100 provide heating, while electricity is used to provide cooling and power the fan.

101 The HVAC system is controlled based on the VAV 2A2-21232 sequence from
 102 the Sequences of Operation for Common HVAC Systems described in [24].

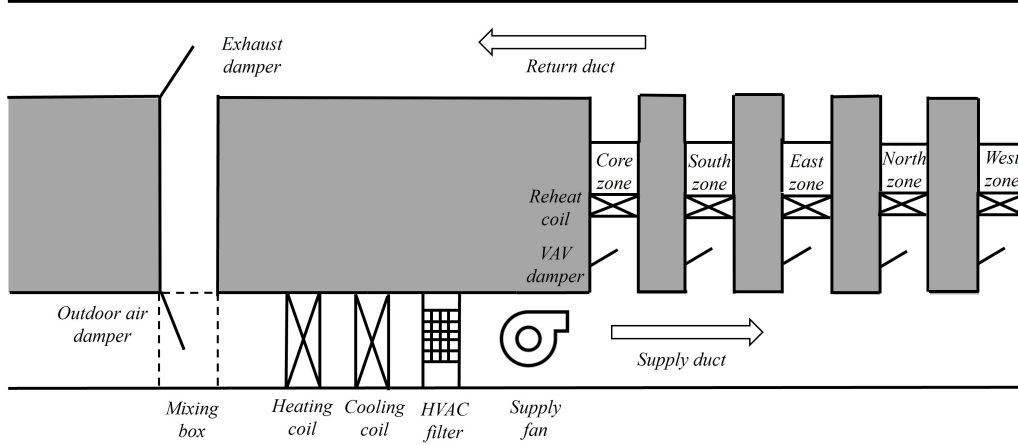


Figure 1: Schematic of VAV system servicing the bottom floor of the five zone medium office building.

103 2.2. System Modeling

104 The five zone, medium office building system model is developed using the
 105 Modelica *Buildings* library for this study. The HVAC system is sized for each
 106 climate using EnergyPlusTM and the fan is assumed to be sized for MERV
 107 10 filtration. We use typical meteorological year data for each location [25].
 108 More about the original building system model can be found in [10].

109 The previous model was designed for a cold and dry climate, so the
 110 air-conditioning (AC) system can be turned off when there are no occupants.

111 However, when the system is used in a humid climate (e.g., Tampa), the
112 AC has to run at all times to avoid development of mold due to the high
113 humidity. For the system located in Tampa in this study, the model is
114 adapted to supply air through the building at all times, including unoccupied
115 hours. The outdoor air damper is closed during unoccupied hours and only
116 recirculated air is supplied to the building (including for the 100% outdoor
117 air case). For cooling scenarios, the supply air temperature setpoint is reset
118 from 12 °C to 27 °C and the zone temperatures are reset from 24 °C to 30 °C
119 in unoccupied hours. For heating scenarios, the zone temperatures are reset
120 from 20 °C to 12 °C in unoccupied hours. This allows for the system to run
121 and prevent buildup of mold, while limiting the increase in energy during the
122 unoccupied hours.

123 The dew point temperature in the core zone for the system in Tampa when
124 the system is always running compared to when the system turns off during
125 unoccupied hours is shown in Figure 2. The two days shown are Sunday and
126 Monday, August 25 and 26. When the system does not run on Sunday, the
127 dew point temperature in this zone increases above the acceptable limit of
128 15 °C (according to ASHRAE Standard 62.1 [23]) for over 8 hours due to
129 infiltration of humid air in the building. After the system turns on Monday

130 morning, the dew point temperature drops back to an acceptable range.
 131 On the other hand, the dew point temperature in this zone remains in an
 132 acceptable range during this time when the system runs 24/7.

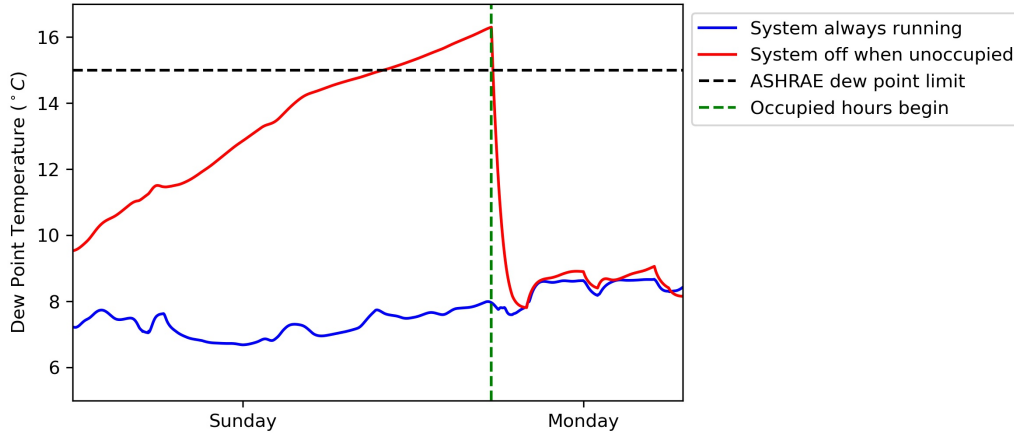


Figure 2: Dew point temperature in the core zone for the system in Tampa on August 25-26.

133 3. Methods to Compare Mitigation Strategies

134 The methods to compare the mitigation strategies in terms of IAQ, financial
 135 costs, and CO₂ emissions are detailed in this section.

136 3.1. Indoor Air Quality Calculation

137 Indoor air quality can consider several factors, such as chemical and
 138 biological compounds, particulates, and gases [26]. To narrow the scope,

139 this study focuses on indoor biological compounds, using the COVID-19
 140 pandemic as a scenario for analysis. Thus, IAQ is represented by the building
 141 level concentration of COVID-19 virus in this study. The sick people generate
 142 viral particles directly into each well-mixed zone at a constant generation
 143 rate. The balance of concentration in a zone can be described as:

$$\frac{dc_{zone}}{dt} = (1/m_{air,zone})[\Sigma(\dot{m}c)_{in} - \Sigma(\dot{m}c)_{out}] + \dot{c}_{gen,zone} - \dot{c}_{decay,zone}, \quad (1)$$

144 where $\frac{dc_{zone}}{dt}$ is the rate of change of virus concentration in the zone with
 145 respect to time, $m_{air,zone}$ is the mass of air in the zone, $\Sigma(\dot{m}c)_{in}$ is the
 146 sum of the virus concentration flowrates into the zone, $\Sigma(\dot{m}c)_{out}$ is the sum
 147 of the virus concentration flowrates out of the zone, $\dot{c}_{gen,zone}$ is the virus
 148 concentration generation rate within the zone, and $\dot{c}_{decay,zone}$ is the rate of
 149 viral decay in the zone, which is modeled based on a first order method:

$$\dot{c}_{decay,zone} = k_{decay}c_{zone}, \quad (2)$$

150 where k_{decay} is a defined constant rate of viral decay, and c_{zone} is the virus
 151 concentration in the zone.

152 We simulate the presence of one sick person in each zone within the
 153 building from 9:00 AM - 5:00 PM, Monday through Friday throughout the

154 year. This allows for the evaluation of the mitigation strategies during
155 different conditions, such as weather, throughout the year. We select a typical
156 virus generation rate of 25 *quanta/hr* [27, 28] and a viral decay rate of 0.48
157 hr^{-1} [6] based on data from the literature. The final results for IAQ presented
158 in Section 5 are calculated based on the average virus concentration in all
159 the zones, averaged over the year during occupied hours.

160 3.2. Financial Cost Calculation

161 The annual financial costs for the different mitigation strategies are calculated
162 based on the following equation:

$$J_{total} = J_{filter} + J_{elec} + J_{gas}, \quad (3)$$

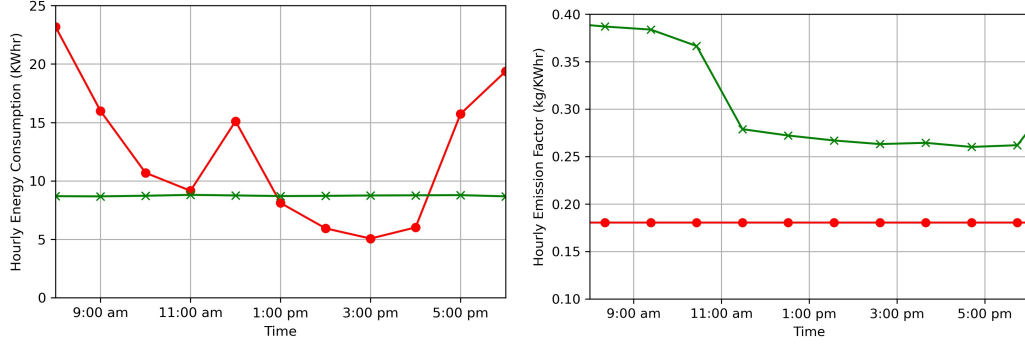
163 where J_{total} is the total annual costs, J_{filter} are the costs associated with
164 filtration, J_{elec} are the electricity costs to run the HVAC system, and J_{gas}
165 are the costs for natural gas heating. The costs associated with filtration
166 include purchase costs and labor costs for replacing the filters throughout
167 the year based on their expected life. The electricity costs to run the HVAC
168 system come from fan and cooling power. Finally, the natural gas costs are
169 calculated based on the heat supplied in the HVAC system from natural gas.

170 3.3. CO_2 Emissions Calculation

171 The annual CO_2 emissions for the mitigation strategies are determined
172 based on emissions associated with natural gas heating and electricity consumed
173 by the HVAC system, using the method adopted in [29, 30]. The emission
174 factor for natural gas heating is constant and independent of location. However,
175 the emission factor for electricity is dynamic and depends on the electricity
176 sources of the location. Different locations use various portions of renewable,
177 nuclear, or fossil fuel energy. The electricity sources vary based on the time
178 of day as well as season, for example depending on the availability of solar
179 or wind energy. The emission factor data comes from the Cambium project
180 lead by the National Renewable Energy Laboratory [31].

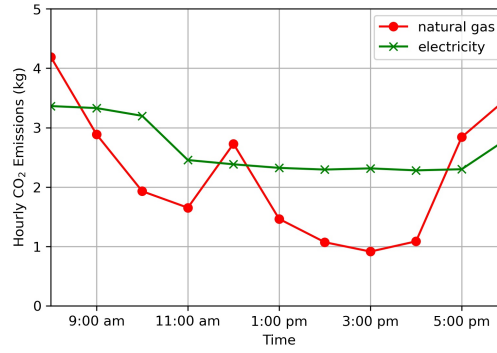
181 Figure 3 shows an example of how CO_2 emissions are calculated for a
182 sample day based on the natural gas and electricity usage. Figure 3a shows
183 the energy consumption for this heating day in Denver. We see the natural
184 gas usage varies based on the heating demand, while the electricity remains
185 constant since only fan power is needed. The emission factor of electricity
186 in Figure 3b varies during the day based on the availability of renewable
187 energy, while the emission factor of natural gas heating remains constant.
188 Finally, Figure 3c shows the hourly CO_2 emissions are the product of the

189 hourly energy usage and emission factor.



(a) Hourly energy consumption.

(b) Hourly emission factors.



(c) Hourly CO₂ emissions.

Figure 3: Calculation of CO₂ emissions based on electricity and natural gas usage for Feb 20, 2020 in Denver.

190 3.4. Analysis of Combined Metrics

191 To evaluate the performance of the different strategies relative to MERV
 192 10, we define a series of metrics by considering the IAQ, costs and/or CO₂
 193 emissions. These are relative metrics, since they are calculated for the

194 strategies relative to MERV 10. First, we calculate the percent increase
 195 in costs or CO₂ emissions relative to MERV 10. This is described as:

$$\Delta J_i = J_i/J_{M10} - 1, \quad (4)$$

196 where ΔJ_i is the percent increase in costs/emissions associated with a strategy
 197 i relative to MERV 10, J_i is the costs/emissions for strategy i , and J_{M10} is
 198 the costs/emissions for MERV 10 in that location.

199 The percent improvement in IAQ relative to the percent increase in
 200 costs/emissions can then be calculated as:

$$\Delta IAQ/\Delta J_i = (1 - IAQ_i/IAQ_{M10})/\Delta J_i, \quad (5)$$

201 where $\Delta IAQ/\Delta J_i$ is the marginal improvement in IAQ per increase in cost/emissions
 202 for a strategy i relative to MERV 10, IAQ_i is the IAQ metric for a strategy
 203 i , and IAQ_{M10} is the IAQ metric for the MERV 10 strategy.

204 We then compare the marginal improvements in IAQ relative to both
 205 costs and emissions by applying a price to CO₂ emissions. We use a cost
 206 of \$12 (USD) per ton of CO₂ emissions based on average prices in the
 207 U.S. described by the Regional Greenhouse Gas Initiative and California
 208 Cap-and-Trade Program [32]. By converting CO₂ emissions to costs, the

209 marginal improvements in IAQ relative to both costs and emissions can be
210 calculated based on Equation 5.

211 4. Scope of Analysis

212 We describe the scope of our analysis in this section, including the selected
213 mitigation strategies, summary of the chosen geographic locations, and list
214 of assumptions.

215 4.1. Mitigation Strategies

216 Four mitigation strategies are chosen for this study, including use of
217 MERV 10, MERV 13, or HEPA filtration, or supply of 100% outdoor air into
218 the building with MERV 10 filtration. The 100% outdoor air strategy also
219 uses MERV 10 filtration, since filtration is needed for outdoor contaminants
220 as well. For brevity, this strategy is referred to simply as “100% outdoor
221 air” in the remainder of this paper. For the cases other than the 100%
222 outdoor air case, the minimum outdoor airflow during occupied hours is
223 defined based on a minimum volumetric outdoor airflow rate, rather than
224 an outdoor air fraction. The outdoor airflow can also increase above the
225 minimum value to provide free cooling based on the outdoor air economizer
226 control. For all cases, including the 100% outdoor air case, the outdoor

227 airflow will only decrease below the minimum value to prevent freezing of
228 the heating/cooling coils. The simulated static pressure drop caused by the
229 HVAC filter varies quadratically with the mass flowrate, as described in [10].
230 It should be noted the pressure drop across the filter can increase over time
231 as the filter accumulates particles [33] and the pressure drop can vary for
232 filters with the same rating, depending on the depth or type of filter [17].
233 For simplicity, a constant nominal pressure drop for each filter is chosen
234 based on the average of the typical initial and final pressure drops. Similarly,
235 the filter particle removal efficiency is dependent on many aspects, such as
236 the size of the particles, loading of filters, and duct leakage [34]. This study
237 assumes the viral particles have diameters between 1-3 μm , and a constant,
238 typical removal efficiency is chosen based on filter data for particles of this
239 size. Table 1 shows the settings for the HVAC filters used in the simulations.
240 The filtration efficiencies come from ASHRAE technical resources [35] and
241 the pressure drop values come from data for MERV 10 [36], MERV 13 [37],
242 and HEPA [38] filters.

Filter	Nominal Pressure Drop (Pa)	Filtration Efficiency
MERV 10	143	50%
MERV 13	162	85%
HEPA	373	99.97%

Table 1: HVAC filter simulation settings.

243 The costs of the HVAC filters, which are obtained from [38], are shown
244 in Table 2. The total annual costs are determined by the purchase and labor
245 costs throughout the year based on the expected life of the filters.

Filter	Purchase Cost (USD)	Labor Costs per Replacement (USD)	Expected Life	Total Annual Costs
MERV 10	\$7	\$17	4 months	\$72
MERV 13	\$11	\$17	4 months	\$84
HEPA	\$150	\$17	12 months	\$167

Table 2: HVAC filter costs.

246 4.2. Geographic Locations and Climates

247 Five unique geographic locations with distinct climates across the United
248 States are selected based on related work [29, 30] to provide a diversity
249 of climates and electricity sources. A summary of the climates, electricity
250 sources, energy prices, and average emission factors from electricity generation
251 is shown in Figure 4. The climates vary from the very cold climate of
252 International Falls, Minnesota to the hot and humid climate of Tampa,
253 Florida. The breakdown of electricity sources in the year 2020 for the five
254 locations [31] are also shown. The average emission factors from electricity
255 generation for each location are included to understand the impact of the
256 electricity sources on CO₂ emissions. Great Falls has the lowest average
257 emission factor since it uses mostly renewable energy from hydropower. San
258 Diego has the second lowest average emission factor, due to utilizing significant
259 renewable energy, such as solar power, and limiting its fossil fuel usage.
260 International Falls, Tampa, and Denver have the highest average emission
261 factors. While Denver and International Falls utilize zero emission sources
262 like wind and nuclear energy, they still rely significantly on fossil fuels like
263 coal and natural gas for electricity. Tampa also heavily relies on fossil fuels,
264 since over 75% of Tampa’s electricity comes from natural gas. It should

265 be noted that electricity sources such as wind and nuclear power have zero
 266 direct emissions, but include emissions when considering the entire life cycle
 267 of production [39, 40]. This study only incorporates direct emissions and not
 268 full life cycle emissions in order to focus on the emissions directly associated
 269 with building operation. The electricity [41] and natural gas [42] prices for
 270 each location are also included. The natural gas price is based on the total
 271 price paid by end-users per thousand cubic feet of natural gas, and is inclusive
 272 of all taxes and other fees.

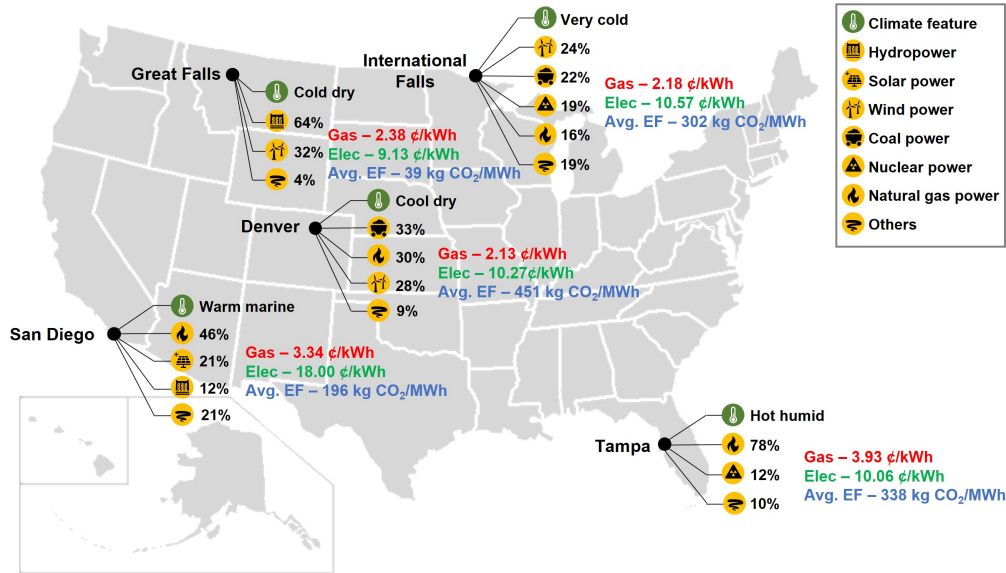


Figure 4: Summary of climate, electricity sources, energy prices, and average emission factor from electricity generation for the five studied locations.

273 4.3. Assumptions

274 The following assumptions are used for this study. First, we assume
275 constant virus generation rates from the sick people and a constant first
276 order viral decay rate value for COVID-19 virus in this work. We assume
277 one sick person per zone working from 9:00 AM to 5:00 PM, Monday through
278 Friday during the entire year. The air in each zone of the office building is
279 also assumed to be well-mixed. We assume constant nominal pressure drop
280 values for each filter, although the actual pressure drop varies based on the
281 airflow rate. The removal efficiencies of the filters are also assumed to be
282 constant. We assume the fan is sized for an existing HVAC system with
283 MERV 10 filtration in all cases. The individual electricity and natural gas
284 prices for each location are constant throughout the year. We use hourly
285 weather data and CO₂ emission data for each location based on the year
286 2020.

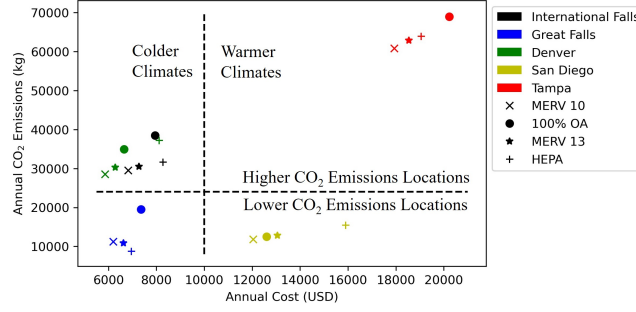
287 5. Results and Discussion

288 We first show an overview of the results for the four mitigation strategies
289 in the five locations in terms of IAQ, financial costs, and CO₂ emissions.
290 We then analyze the results based on the impacts of climate and electricity

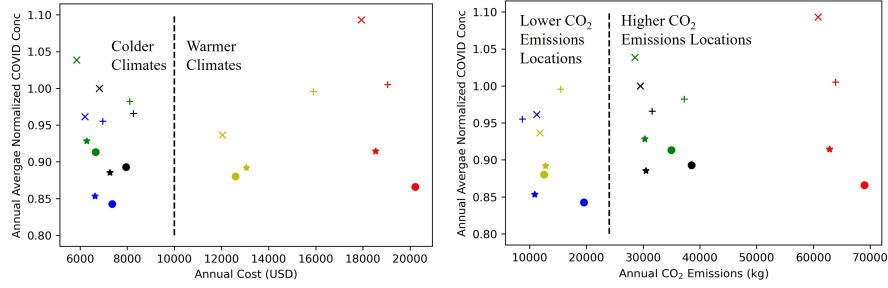
291 sources. Finally, we discuss the results based on the tradeoffs among different
292 user priorities.

293 *5.1. Overview of Results*

294 The annual results for IAQ, costs, and CO₂ emissions are shown in Figure
295 5. The virus concentrations are normalized by the annual average virus
296 concentration for the MERV 10 case in International Falls ($0.011 \text{ quanta}/m^3$).
297 One general result is that HEPA filtration never provides the best IAQ for
298 a given location, and is also always worse than the less efficient MERV
299 13 filtration. In the five locations, MERV 13 filtration improves the IAQ
300 by 5.4-10.6% compared to HEPA filtration. This is because the system is
301 not sized for the additional pressure drop caused by HEPA filtration, which
302 results in reduced overall system flowrates and lower virus removal rates.



(a) Annual CO₂ emissions vs cost.



(b) Average virus concentration vs annual cost. (c) Average virus concentration vs annual CO₂ emissions.

Figure 5: Results for average virus concentration, annual cost, and annual CO₂ emissions for the four mitigation strategies and five locations.

303 The annual results show dependencies on climate and electricity sources.
 304 This can especially be seen in Figure 5a, where the colder climates have
 305 lower annual costs compared to the warmer climates, and there is also a
 306 clear divide between the locations with higher or lower CO₂ emissions from
 307 electricity generation. Figures 5b and 5c similarly show these divides based
 308 on climate and electricity sources, as well as the IAQ trends for the different

309 mitigation strategies. The 100% outdoor air strategy usually provides the
310 best IAQ, but can lead to significant increases in costs and CO₂ emissions.
311 MERV 10 filtration is typically the cheapest and lowest emission strategy,
312 but also usually provides the worst IAQ. MERV 13 filtration improves the
313 IAQ relative to MERV 10 filtration, but with moderate increases in costs
314 and emissions. Finally, HEPA filtration often improves the IAQ relative to
315 MERV 10 filtration, but not compared to MERV 13 filtration or use of 100%
316 outdoor air. It also can lead to significant increases in costs and emissions.
317 Based on these findings, we analyze the impacts of climate and electricity
318 sources in the following subsections.

319 *5.2. Impact of Climate*

320 We discuss the results for the four mitigation strategies in this section
321 based on colder and warmer climates. The colder climates are International
322 Falls, Great Falls, and Denver, while the warmer climates are San Diego and
323 Tampa.

324 *5.2.1. Colder Climates*

325 There are several common trends among the colder climates. International
326 Falls is used as an example in this section, and the breakdown of the results

327 in this location is shown in Figure 6. The key feature of the colder climates
328 is the dominant energy consumption of natural gas for heating. Figure 6a
329 shows, for most cases, the majority of annual energy comes from natural gas
330 heating, especially for the 100% outdoor air case. Despite the significant
331 natural gas usage, Figure 6b shows the costs from natural gas are relatively
332 small compared to those from electricity (used for cooling and fan energy).
333 The percentage of costs associated with natural gas heating range from
334 14-33% for the four cases in this location. This is because natural gas is
335 significantly cheaper than electricity, which is true in all the studied locations.
336 The majority of emissions comes from electricity usage for most of the cases
337 in this location, as shown in Figure 6c. The exception is the 100% outdoor
338 air case, which results in 56% of emissions from natural gas heating due to
339 the energy needed to heat the cold outdoor air.

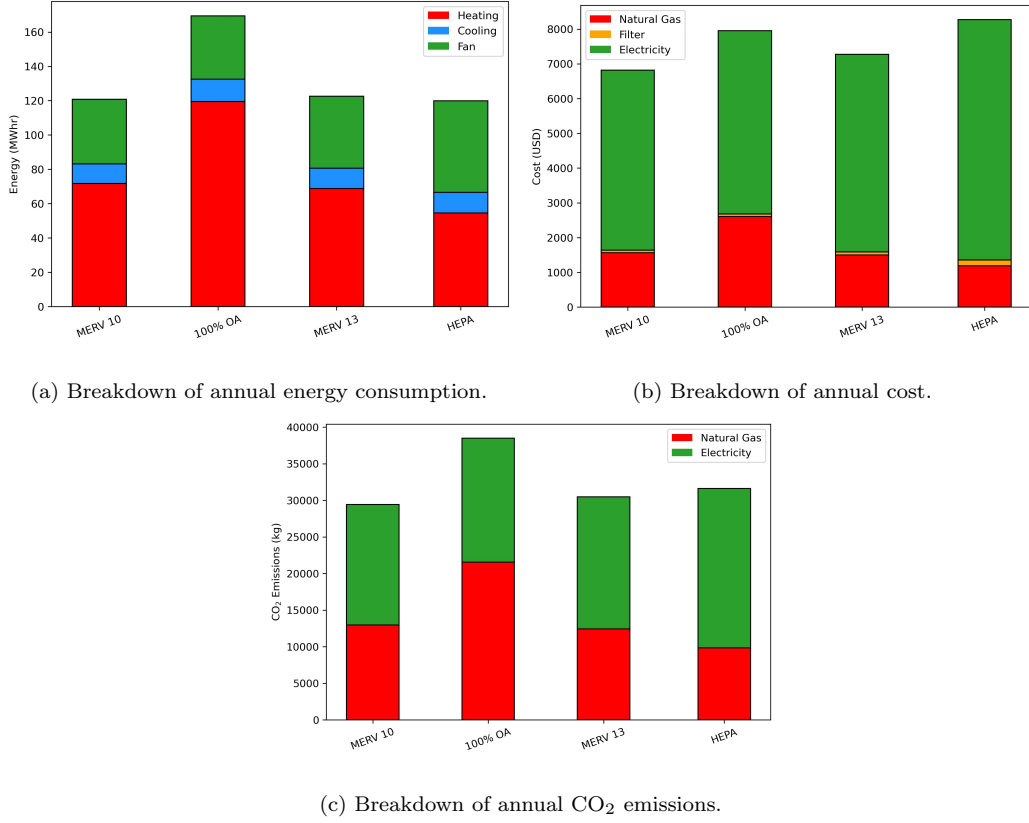


Figure 6: Annual energy, cost, and CO₂ emission results for International Falls.

340 There is also a tradeoff between heating and fan energy for the more
 341 efficient filter cases. The higher pressure drop filters require more fan power
 342 to supply airflow, which results in the fan dissipating more heat to the airflow
 343 as it works harder. This causes the more efficient filter cases to save on some
 344 heating energy, which is especially seen by the HEPA case in Figure 6a. For
 345 the colder climates, the additional heat produced by the fan can be beneficial

346 to efficiently add heat to the system, while not requiring much more cooling
347 energy, since these climates do not require significant cooling. However,
348 this increase in electrical heating leads to higher costs due to the relative
349 price of electricity compared to natural gas heating. It can also increase or
350 reduce emissions depending on the electricity sources in a particular location.
351 Since International Falls uses significant fossil fuel energy in their electricity
352 generation, the more efficient filter cases lead to higher emissions relative to
353 the MERV 10 case.

354 The very cold climate also affects the control of the outdoor air economizer.
355 For the 100% outdoor air case, the economizer will always supply 100%
356 outdoor air (or at least the minimum outdoor airflow for the other cases),
357 except when the outdoor air needs to be reduced to prevent freezing of the
358 coils in the air handling unit. This becomes noticeable for the colder climates.
359 For example, Figure 5c shows that MERV 13 filtration provides better IAQ
360 compared to supply of 100% outdoor air for International Falls, which is
361 not the case for the other locations. This is because the outdoor airflow
362 needs to be reduced often throughout the year to prevent freezing, so MERV
363 13 filtration becomes more effective. Figure 7 shows the dynamic usage of
364 outdoor air throughout the year in International Falls for the 100% outdoor

365 air case. We see this strategy can supply 100% outdoor air in the warmer
 366 months, but often has to reduce the outdoor airflow in the winter and colder
 367 mornings. As a result, 100% outdoor air is only supplied about 55% of the
 368 time during occupied hours.

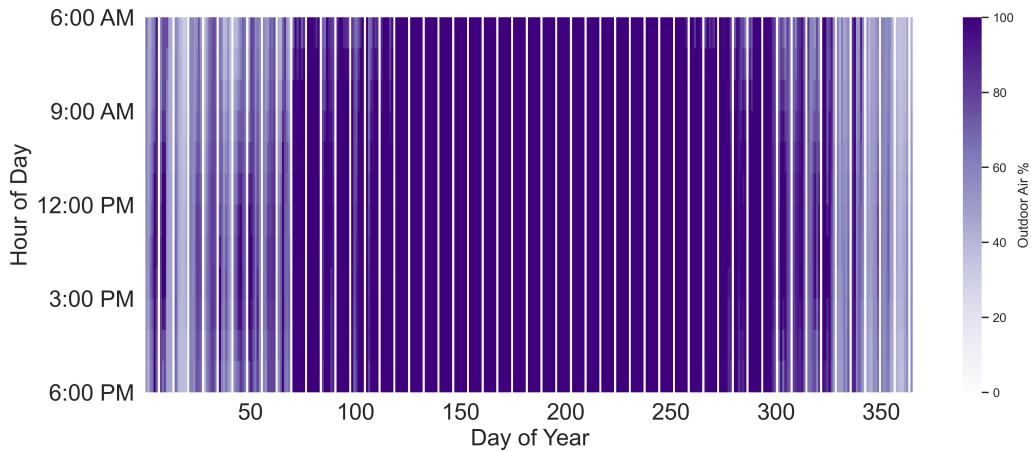
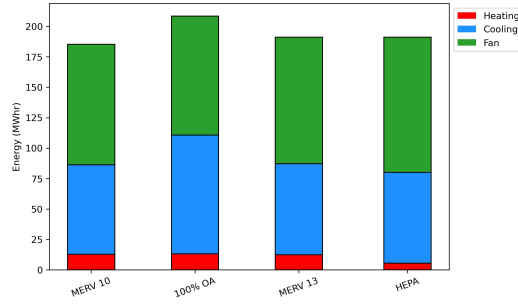


Figure 7: Dynamic usage of outdoor air in International Falls for the 100% outdoor air strategy.

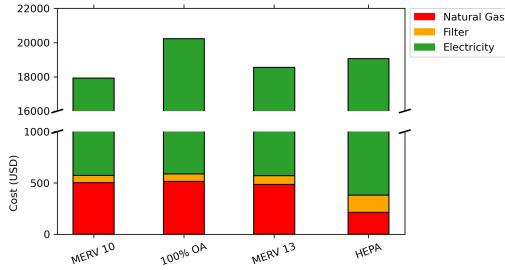
369 5.2.2. Warmer Climates

370 Next, there are some typical trends in the warmer climates. Compared
 371 to the colder climates, which use a lot of natural gas for heating, the warmer
 372 climates rely heavily on electricity for cooling and use very little natural gas.
 373 As an example of a warmer climate, Figure 8 shows the results for Tampa,
 374 which is considered a hot and humid climate by ASHRAE. The low usage

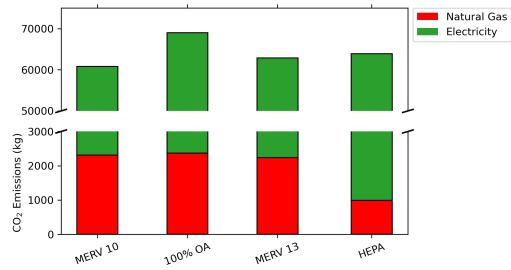
375 of natural gas heating leads to much lower costs and emissions from natural
 376 gas compared to higher costs and emissions from electricity. Less than 3%
 377 of the costs and 4% of the emissions come from natural gas heating for the
 378 four strategies in this location. The relative price of electricity compared to
 379 natural gas and reliance on electricity in warmer climates is the reason for
 380 the higher annual costs in the warmer climates, as seen in Figure 5a.



(a) Breakdown of annual energy consumption.



(b) Breakdown of annual cost.

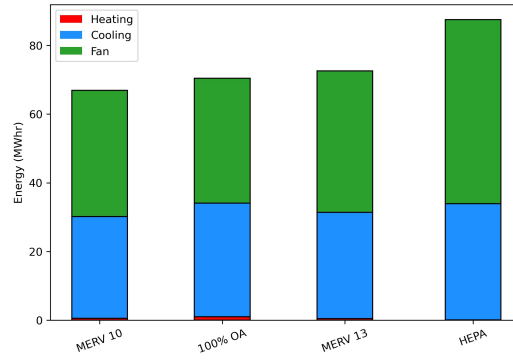


(c) Breakdown of annual CO₂ emissions.

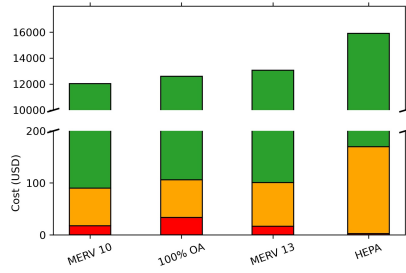
Figure 8: Annual energy, cost and CO₂ emission results for Tampa.

381 For Tampa, use of 100% outdoor air leads to a 33% increase in cooling
 382 energy (including dehumidification) relative to MERV 10 filtration because of

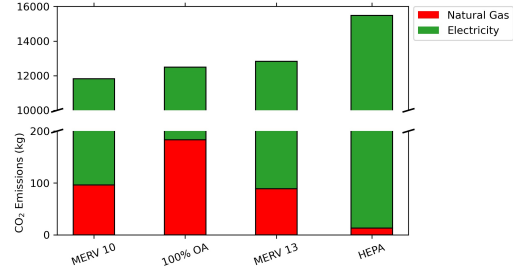
both the heat and humidity in this climate. This also leads to large increases in costs and emissions. In San Diego, however, supplying 100% outdoor air does not increase the costs and emissions as much, as seen in Figure 9. This is due to the relatively milder weather and lower humidity compared to Tampa.



(a) Breakdown of annual energy consumption.



(b) Breakdown of annual cost.

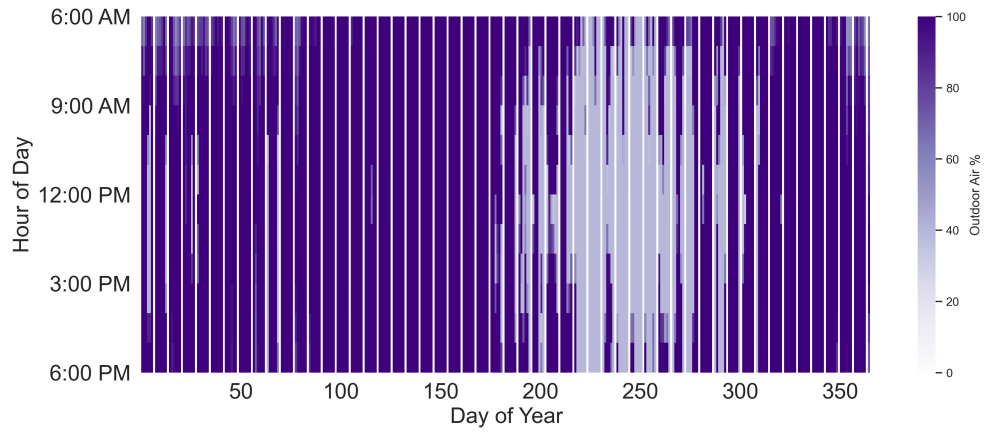


(c) Breakdown of annual CO₂ emissions.

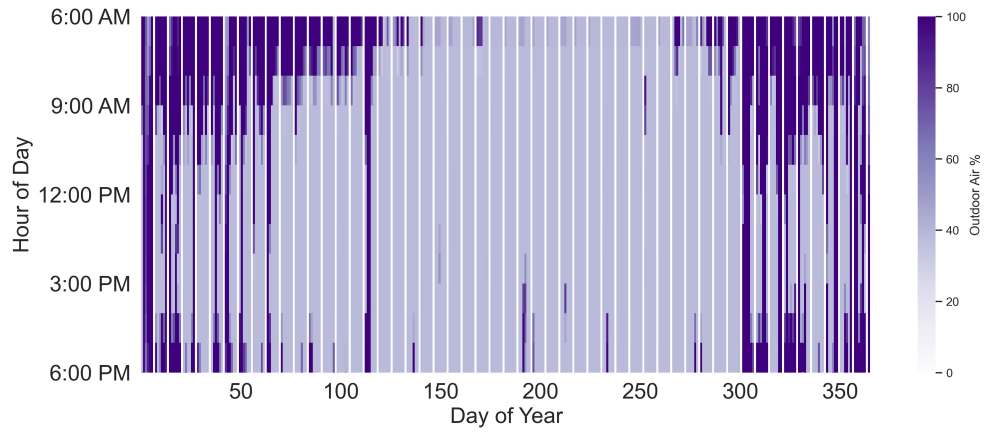
Figure 9: Annual energy, cost and CO₂ emission results for San Diego.

This weather in San Diego also allows for more outdoor air use for the filter cases using the airside economizer throughout the year, which affects the virus concentration results shown in Figure 5b. There are relatively

390 smaller differences among the virus concentrations for the MERV 10, MERV
391 13, and 100% outdoor air cases due to the high outdoor air usage in San
392 Diego. MERV 10 filtration even improves the IAQ by 6% compared to HEPA
393 filtration due to the significant amount of outdoor air supplied for this climate
394 and the reduced flowrates caused by the high pressure drop of the HEPA
395 filter. Figure 10 shows the dynamic outdoor air usage throughout the year
396 for the MERV 10 cases in San Diego and Tampa. This shows the high usage of
397 outdoor air in San Diego due to its milder weather, although less outdoor air
398 is used during the hotter months from July through October. For reference,
399 the monthly average outdoor temperatures for these two locations are shown
400 in Figure 11. In comparison, not much outdoor air is used for the filter
401 cases in Tampa due to the heat and high humidity, as shown in Figure 10b.
402 This leads to larger differences in virus concentrations among the MERV 10,
403 MERV 13, and 100% outdoor air cases for Tampa as seen in Figure 5b.



(a) San Diego (warm and marine).



(b) Tampa (hot and humid).

Figure 10: Dynamic usage of outdoor air using MERV 10 filtration in San Diego and Tampa.

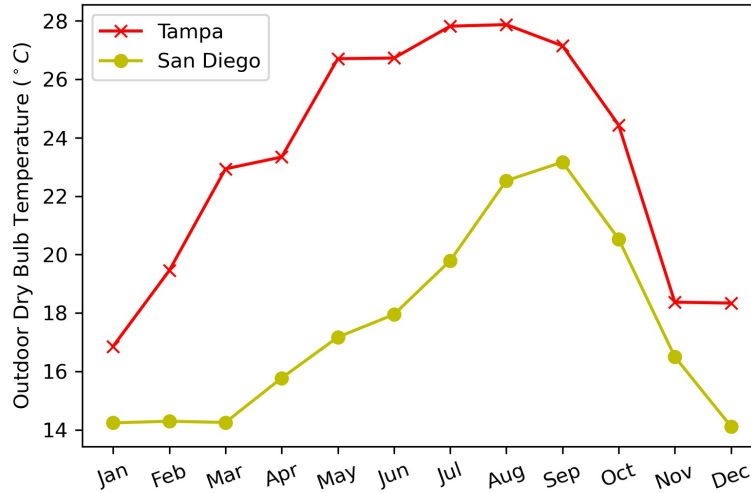


Figure 11: Average monthly temperatures in Tampa and San Diego.

404 Finally, the increased heat dissipated by the fan for the more efficient
 405 filter cases is more penalizing for the warmer climates. Compared to the
 406 colder climates, the additional heat from the fan is not typically needed and
 407 rather requires the system to provide more cooling. This leads to higher costs
 408 and emissions for the more efficient filter cases relative to MERV 10, as seen
 409 in Figures 8 and 9.

410 5.3. Impact of Electricity Sources

411 Next, the impact of electricity sources on the results are analyzed in this
 412 section. Great Falls and San Diego are the locations with lower CO₂ emissions
 413 from electricity, while International Falls, Denver, and Tampa have higher

414 CO₂ emissions from electricity. About 96% of the electricity generation in
415 Great Falls comes from the renewable sources of hydro and wind power,
416 making it the lowest emissions from electricity location in this study. San
417 Diego limits its fossil fuel usage while utilizing significant renewable energy.
418 International Falls, Denver, and Tampa rely heavily on fossil fuels like coal
419 and natural gas for electricity generation.

420 *5.3.1. Locations with Low CO₂ Emissions from Electricity Generation*

421 First, we present results for the cleaner electricity locations, using Great
422 Falls as an example. The dynamic CO₂ emission factor from electricity
423 throughout the year for Great Falls is shown in Figure 12. The emission
424 factor for electricity exceeds the emission factor for natural gas heating
425 (180 kg/MWhr) during only about 11% of the year. It often utilizes 100%
426 renewable energy for electricity resulting in an emission factor of zero, and
427 has an average emission factor throughout the year of about 39 kg/MWhr.

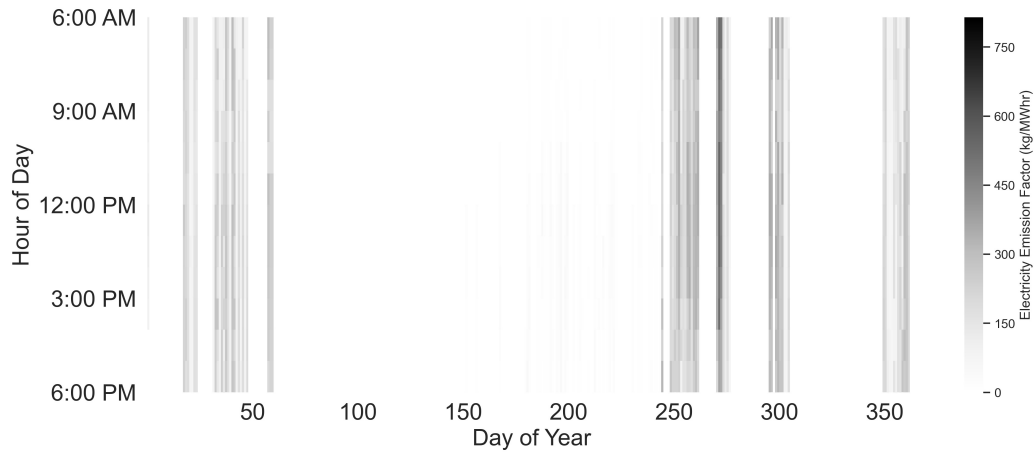
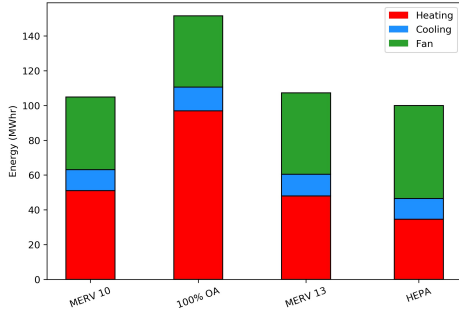
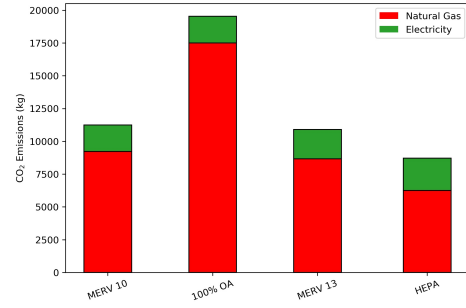


Figure 12: Dynamic CO₂ emission factor from electricity in Great Falls.

Figure 13 shows the breakdown of energy consumption and CO₂ emissions for this location. Unlike the similarly cold climate of International Falls, its electricity largely comes from clean hydropower. Thus, its emissions mainly come from natural gas heating rather than electricity. In this case, even for the highest emission scenario of using 100% outdoor air, a building in Great Falls will produce less emissions than one in the other studied cold climates. For example, use of 100% outdoor air in Great Falls produces about 32% less emissions than MERV 10 filtration in Denver.



(a) Breakdown of annual energy consumption.



(b) Breakdown of annual CO₂ emissions.

Figure 13: Results for Great Falls.

Furthermore, the additional electrical heating dissipated by the fan in the
 efficient filter cases leads to a further reduction in emissions for these cases
 when the electricity is coming from low emissions sources, as seen in Figure
 13b. This is because the small increase in emissions from electricity to power
 the fan for these cases is offset by the reduction in emissions from natural
 gas heating due to the heat added by the fan. Thus, HEPA filtration has
 the lowest emissions in this location, when it typically has one of the highest
 emissions in other locations.

5.3.2. Locations with High CO₂ Emissions from Electricity Generation

Next are the results for the high CO₂ emissions from electricity locations.
 The results for energy consumption and CO₂ emissions in Denver are shown
 as an example in Figure 14. Despite a significant portion of energy consumption

448 from natural gas heating, especially with the 100% outdoor air case, Figure
 449 14a shows the majority of emissions comes from electricity.

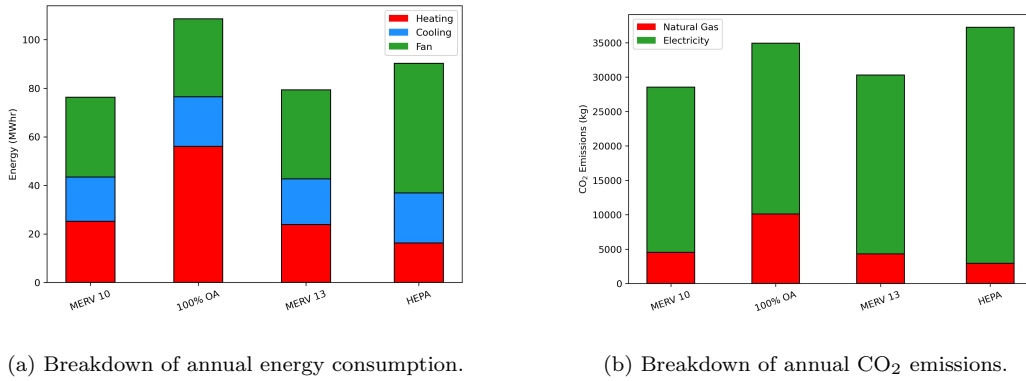


Figure 14: Results for Denver.

450 The high emissions from electricity is because the electricity generation
 451 in Denver mainly comes from burning fossil fuels such as coal and natural
 452 gas. Thus, despite the 100% outdoor air case using more energy than the
 453 HEPA case, the HEPA case results in more emissions due to the electricity
 454 usage over natural gas. The dynamic CO₂ emission factor from electricity
 455 throughout the year in Denver is shown in Figure 15. The emission factor
 456 from electricity exceeds that from natural gas heating about 99% of the time
 457 in Denver. Because of this, the increase in electricity and decrease in heating
 458 caused by the higher fan power for the more efficient filter cases further
 459 increases the emissions for these cases, as shown in Figure 14.

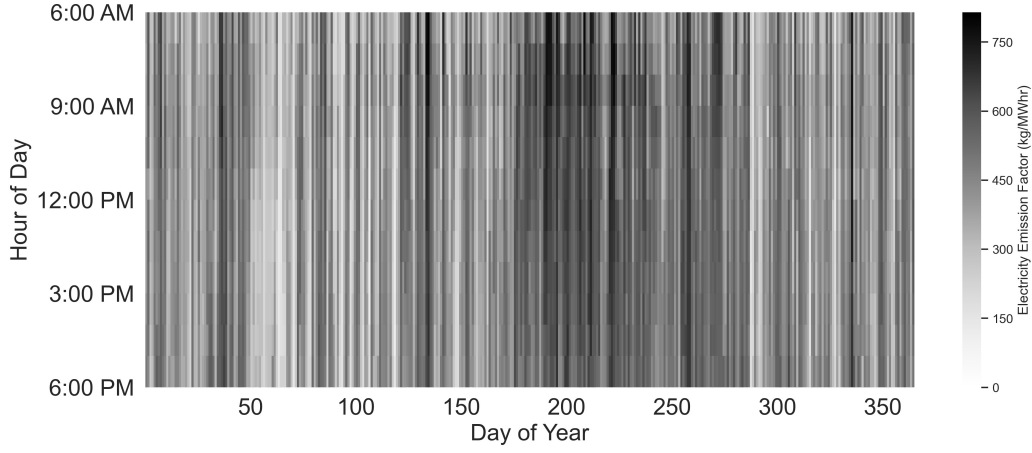


Figure 15: Dynamic CO₂ emission factor from electricity in Denver.

460 5.4. Findings Based on Priority

461 The results based on user priority are summarized in this section. For
 462 each climate, the strategies can be compared relative to MERV 10 filtration
 463 based on the metrics of IAQ, costs, and CO₂ emissions, or any combination
 464 of these metrics. We first present the results based on a single priority, then
 465 analyze the results with a combination of priorities.

466 5.4.1. Results for Individual Priorities

467 For each strategy, the results for IAQ, costs, and CO₂ emissions are
 468 normalized by the results using MERV 10 filtration in that location. Thus,
 469 the MERV 10 results are always equal to one since they are normalized by
 470 themselves. A number less than one represents an improvement relative

471 to MERV 10, signifying a reduction in indoor virus concentration, costs,
 472 or emissions. Conversely, a number greater than one represents a worse
 473 performance relative to MERV 10, such as an increase in indoor virus concentration,
 474 costs, or emissions. The results relative to MERV 10 filtration are shown for
 475 International Falls in Table 3, and similar tables for the remaining locations
 476 are included in the appendix.

Strategy	IAQ	Cost	CO ₂
MERV 10	1	1	1
100% OA	0.89	1.17	1.31
MERV 13	0.89	1.07	1.03
HEPA	0.97	1.21	1.07

Table 3: Results for the strategies relative to MERV 10 for the individual metrics in International Falls.

477 There are trends for the best strategy based on a single priority for the
 478 different locations. In four of the five locations, supply of 100% outdoor air
 479 provides the best IAQ. The exception occurs in International Falls, whose
 480 very cold climate prevents the use of 100% outdoor air during the coldest
 481 times of the year to avoid freezing of the coils in the air handling unit.

482 MERV 13 filtration provides the second best IAQ in all locations, except
483 International Falls, where it has slightly better IAQ compared to 100%
484 outdoor air. HEPA filtration is usually third best for IAQ due to the reduced
485 flowrates caused by the high pressure drop of the filter, although its high
486 particle removal efficiency usually allows it to outperform MERV 10 filtration.
487 MERV 10 filtration provides the worst IAQ in all locations except San Diego,
488 where the high outdoor air usage allows it to outperform HEPA filtration.
489 Based on these results, there are tradeoffs between filter efficiency and pressure
490 drop (and resulting airflow rate). There should be a theoretical ideal balance
491 between filter efficiency and pressure drop, which would likely be dependent
492 on many factors including climate. In this study, the differences in airflow
493 rates become very important for the efficient filters and our findings show a
494 slightly less efficient filter with significantly lower pressure drop is preferable.

495 MERV 10 filtration has the lowest costs in all five locations due to its low
496 energy usage compared to the other cases. In four of the five locations, MERV
497 13 filtration has the second lowest costs. The exception is in San Diego,
498 where 100% outdoor air has lower costs since the milder weather causes a
499 smaller increase in heating/cooling energy for 100% outdoor air relative to
500 the increase in electricity to power the fan for the MERV 13 case. Use of

100% outdoor air in Tampa, Great Falls, and International Falls leads to the highest costs in these locations due to the more extreme weather. Finally, use of HEPA filtration leads to the highest costs in Denver and San Diego, where the costs from the increased fan power for the HEPA case outweighs the increase in costs for 100% outdoor air. These two locations also have relatively milder weather compared to the other locations, which explains why the increase in costs from 100% outdoor air is less significant.

MERV 10 filtration also has the lowest CO₂ emissions in four of the five locations. Similar to having the lowest costs, this is because MERV 10 filtration tends to use the least energy. The exception is in Great Falls, where the reduced natural gas heating for the efficient filter cases caused by the increased heat dissipated by the fan leads to lower overall emissions. This is because of the high use of renewable energy in Great Falls, so the small increase in emissions from electricity are offset by the reduction in emissions from natural gas heating for the efficient filter cases. For Great Falls, the rank of CO₂ emissions from lowest to highest is: 1) HEPA, 2) MERV 13, 3) MERV 10, and 4) 100% outdoor air. The 100% outdoor air strategy has the highest CO₂ emissions in International Falls, Great Falls, and Tampa. These are the climates with the most extreme weather, so use of 100% outdoor air

520 results in higher emissions from increased heating/cooling. HEPA filtration
521 results in the highest emissions in Denver and San Diego due to the increase
522 in electricity consumption. The weather in these climates is milder compared
523 to the others, so use of 100% outdoor does not result in as high emissions
524 compared to HEPA filtration. Finally, MERV 13 filtration typically has the
525 second or third lowest CO₂ emissions due to its moderate energy usage.

526 *5.4.2. Combination of Priorities*

527 An optimal strategy can be selected for user's with a combination of
528 priorities as well. Figure 16 shows the comparison of the marginal improvement
529 in IAQ per increase in emissions vs the marginal improvement in IAQ per
530 increase in costs for the different strategies relative to MERV 10 in the five
531 locations.

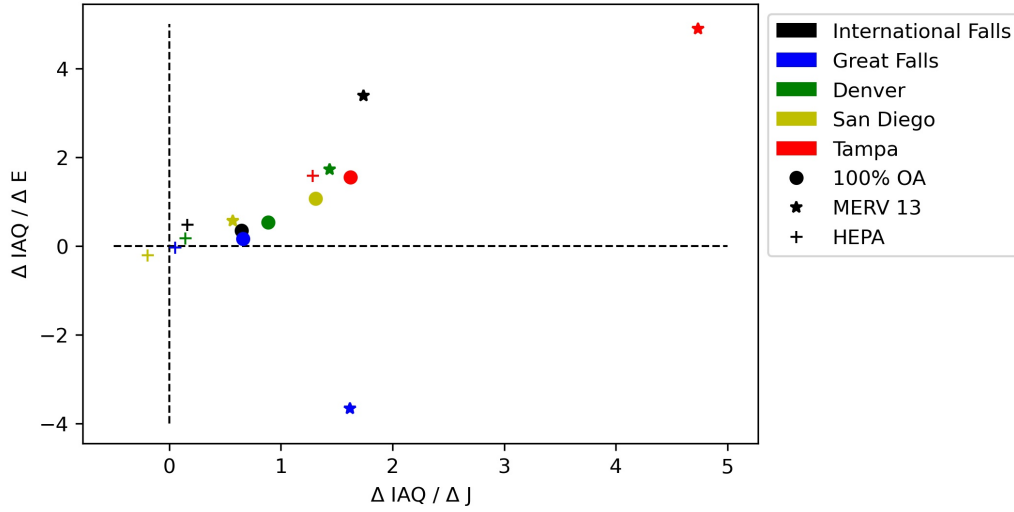


Figure 16: Marginal improvement in IAQ relative to costs and CO₂ emissions for the five locations.

Based on the method to calculate these metrics (described in Section 3.4), a higher positive number for these metrics means the strategy is more beneficial. For example, it represents a greater improvement in IAQ with a smaller increase in costs or emissions relative to MERV 10. Thus, the markers in the upper right hand corner perform the best in terms of improvement in IAQ relative to both costs and emissions. MERV 13 filtration in International Falls and Tampa are the best examples for this, since they can greatly improve the IAQ with limited increases in costs and emissions in these locations. The more extreme weather in these climates means the MERV

10 cases use less outdoor air throughout the year, and the 100% outdoor air cases result in more significant penalties in terms of costs and emissions, making MERV 13 filtration a good option. MERV 13 filtration also performs the best for Denver, although its improvement relative to 100% outdoor air is not as significant as the previously mentioned locations. Use of 100% outdoor air performs the best for San Diego because of its milder weather, resulting in less of a penalty in terms of costs and emissions for this case.

While both these metrics are usually positive, there are three cases where they become negative, two of which occur in Great Falls. The metrics are typically positive due to the sign convention of the calculations: an improvement in IAQ relative to MERV 10 is positive and an increase in costs/emissions relative to MERV 10 is positive. However, the reduction in emissions for the MERV 13 and HEPA cases relative to MERV 10 in Great Falls causes $\Delta IAQ/\Delta E$ to be negative for these cases. In this case, the negative sign represents a more beneficial strategy, for example MERV 13 filtration in Great Falls results in a significant improvement in IAQ with a small improvement in emissions relative to MERV 10. Similarly, the HEPA case sees a small improvement in IAQ with a more significant reduction in emissions relative to MERV 10. The final case with negative values is the

560 HEPA case in San Diego. HEPA filtration results in worse IAQ relative to
561 MERV 10 in San Diego because of the high outdoor air usage for MERV 10
562 in this climate and reduced flowrates for the HEPA filter case. In this case,
563 the negative sign represents a non-beneficial strategy, because it worsened
564 the IAQ and increased the costs and emissions relative to MERV 10.

565 Finally, associating a cost with CO₂ emissions allows us to directly compare
566 the marginal improvement in IAQ to both these metrics simultaneously. This
567 is shown for the three strategies relative to MERV 10 in the five locations in
568 Figure 17.

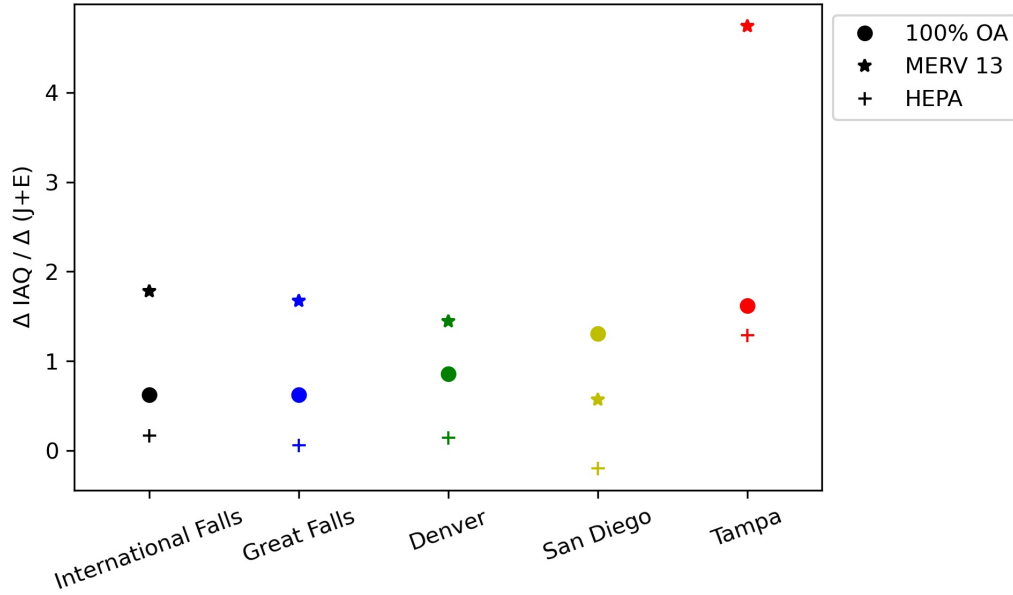


Figure 17: Marginal improvement in IAQ relative to costs including the cost of CO₂ emissions for the five locations.

569 MERV 13 filtration appears to be the most beneficial strategy in four of
 570 the five locations. As seen before, 100% outdoor air is able to outperform
 571 MERV 13 filtration in San Diego due to the milder weather. MERV 13
 572 filtration shows the greatest improvement in Tampa due to the limited outdoor
 573 air usage for the MERV 10 case and the significant penalty in costs for the
 574 100% outdoor air case. HEPA filtration is the least beneficial strategy for
 575 all the climates due to the small increase in IAQ relative to high increases
 576 in costs. For this metric, the only negative number occurs for the HEPA

577 case in San Diego, since HEPA filtration worsens the IAQ relative to MERV
578 10. We do not see the negative numbers for the Great Falls cases since the
579 reduction in emissions is offset by the increase in other costs for the MERV
580 13 and HEPA cases.

581 **6. Conclusion**

582 The tradeoffs among IAQ, financial costs, and CO₂ emissions for four
583 strategies to mitigate indoor virus are compared for five locations across the
584 United States. The mitigation strategies include different levels of filtration,
585 such as MERV 10, MERV 13, or HEPA filtration, as well as supply of 100%
586 outdoor air into the building. The locations have a variety of climates
587 ranging from very cold to hot and humid. Their electricity profiles are
588 also comprised differently, with varying portions of renewable energies and
589 fossil fuels for generating electricity. The strategies are evaluated using
590 a prototypical medium office building model initially sized for MERV 10
591 filtration, developed using the Modelica *Buildings* library.

592 The results show the best solution is dependent on climate, electricity
593 profile, and user priority. MERV 10 filtration is often the best option when
594 the user cares most about costs and/or CO₂ emissions, since this strategy

595 tends to use the least energy. Use of 100% outdoor air usually provides the
596 best IAQ, although often significantly increases costs and CO₂ emissions.
597 The results show this can be a good option in the relatively milder climate
598 of San Diego, where the increase in costs and emissions is limited. MERV 13
599 filtration can provide a nice balance of the three metrics in most locations
600 due to its virus filtration efficiency and relatively smaller increases in energy
601 consumption. This strategy outperforms 100% outdoor air in the locations
602 with more extreme weather, since it avoids the significant increase in heating/cooling
603 outdoor air in these locations. Finally, HEPA filtration should be avoided
604 for this system, and similar systems that are not sized to overcome the high
605 pressure drops of these filters. This leads to large increases in fan power and
606 reductions in system flowrates, leading to high costs and emissions with little
607 improvement in IAQ.

608 Future studies can be conducted based on the work in this paper. The
609 models we used in this study can be applied to other contaminant scenarios,
610 for example PM_{2.5} which can infiltrate the building from outdoor air. Other
611 indoor contaminants can be considered as well, such as CO₂, which can
612 affect worker productivity [43] and quality of sleep [44]. They can also
613 be used to evaluate advanced control strategies to improve IAQ, such as

614 occupant-based strategies. We can also study tradeoffs among energy, costs,
615 and CO₂ emissions for other indoor virus mitigation strategies, such as use of
616 portable air cleaners, which have been shown to be effective at reducing virus
617 concentrations within rooms [45]. Finally, this study focuses on applying
618 mitigation strategies to an existing building, since redesigning an HVAC
619 system is costly. However, the models can be used to evaluate HVAC system
620 designs for new buildings, for example to study a system designed for HEPA
621 filtration.

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631 **Appendix A**

Location	Strategy	Normalized IAQ	Costs (USD)	CO ₂ Emissions (kg CO ₂)
International Falls	MERV 10	1.00	6820	29500
	100% OA	0.89	7950	38500
	MERV 13	0.89	7270	30500
	HEPA	0.97	8270	31600
Great Falls	MERV 10	0.96	6201	11242
	100% OA	0.84	7365	19528
	MERV 13	0.85	6632	10896
	HEPA	0.95	6962	8717
Denver	MERV 10	1.04	5856	28555
	100% OA	0.91	6654	34943
	MERV 13	0.93	6290	30314
	HEPA	0.98	8113	37235
San Diego	MERV 10	0.94	12041	11825
	100% OA	0.88	12597	12493
	MERV 13	0.89	13061	12823
	HEPA	1.00	15902	15473
Tampa	MERV 10	1.09	17928	60829
	100% OA	0.87	20224	69007
	MERV 13	0.91	18548	62861
	HEPA	1.01	19054	63919

Table 4: Results for the four strategies in all five climates for IAQ, costs, and CO₂ emissions.

632 **Appendix B**

Strategy	IAQ	Cost	CO ₂
MERV 10	1	1	1
100% OA	0.88	1.19	1.74
MERV 13	0.89	1.07	0.97
HEPA	0.99	1.12	0.78

Table 5: Results for the strategies relative to MERV 10 for the individual metrics in Great Falls.

Strategy	IAQ	Cost	CO ₂
MERV 10	1	1	1
100% OA	0.88	1.14	1.22
MERV 13	0.89	1.07	1.06
HEPA	0.95	1.39	1.30

Table 6: Results for the strategies relative to MERV 10 for the individual metrics in Denver.

Strategy	IAQ	Cost	CO ₂
MERV 10	1	1	1
100% OA	0.94	1.05	1.06
MERV 13	0.95	1.08	1.08
HEPA	1.06	1.32	1.31

Table 7: Results for the strategies relative to MERV 10 for the individual metrics in San Diego.

Strategy	IAQ	Cost	CO ₂
MERV 10	1	1	1
100% OA	0.80	1.13	1.13
MERV 13	0.84	1.03	1.03
HEPA	0.92	1.06	1.05

Table 8: Results for the strategies relative to MERV 10 for the individual metrics in Tampa.

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