

Cool roofs could be most effective at reducing outdoor urban temperatures in London compared with other roof top and vegetation interventions: a mesoscale urban climate modelling study

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September 25, 2023

Abstract

Comprehensive studies comparing impacts of building and street levels interventions on air temperature at metropolitan scales are still lacking despite increased urban heat-related mortality and morbidity. We therefore model the impact of 9 interventions on air temperatures at 2 m during 2 hot days from the summer 2018 in the Greater London Authority area using the WRF BEP-BEM climate model. We find that on average cool roofs most effectively reduce temperatures ($\sim -1.2^{\circ}\text{C}$), outperforming green roofs ($\sim 0^{\circ}\text{C}$), solar panels ($\sim -0.3^{\circ}\text{C}$) and street level vegetation ($\sim -0.3^{\circ}\text{C}$). Application of air conditioning across London increase air temperatures by $\sim +0.15^{\circ}\text{C}$ but related energetic consumption could be covered by energy production from solar panels. Current realistic deployments of green roofs and solar panels are ineffective at large scale reduction of temperatures. We provide a detailed decomposition of the surface energy balance to explain changes in air temperature and guide future decision-making.

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11 **Key Points:**

- 12 • City scale deployment of cool roofs have the highest impact on 2 m air temper-
13 ature
14 • Green roofs do not decrease daily average temperature but have a daytime cool-
15 ing effect
16 • Solar PV can reduce temperatures in London by capturing sensible heat, gener-
17 ating electrical power and reducing sensible heat flux

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Abstract

Comprehensive studies comparing impacts of building and street level interventions on air temperature at metropolitan scales are still lacking despite increased urban heat-related mortality and morbidity. We therefore model the impact of 9 interventions on air temperatures at 2 m during 2 hot days from the summer 2018 in the Greater London Authority area using the WRF BEP-BEM climate model. We find that on average cool roofs most effectively reduce temperatures (~ -1.2 °C), outperforming green roofs (~ 0 °C), solar panels (~ -0.3 °C) and street level vegetation (~ -0.3 °C). Application of air conditioning across London increase air temperatures by $\sim +0.15$ °C but related energetic consumption could be covered by energy production from solar panels. Current realistic deployments of green roofs and solar panels are ineffective at large scale reduction of temperatures. We provide a detailed decomposition of the surface energy balance to explain changes in air temperature and guide future decision-making.

Plain Language Summary

Multiple common city scale passive and active interventions exist to reduce urban population's exposure to extreme heat during hot spells. Nonetheless, a proper comparison of the effect that each of these interventions may have on the temperatures experienced within large cities is missing. Additionally, energetic mechanisms that lead to outdoor temperature changes are often not detailed and could lead to detrimental effects for local populations, such as indirect increase of relative humidity or outdoor temperatures. Our study, focusing over London, compares several common interventions through a modelling experiment and finds that cool roofs largely outperform other interventions during the two hottest days of the summer 2018. We also find that green roofs are ineffective on average and that solar panels and tree vegetation would only marginally change temperature exposures. Large scale deployment of air conditioning would lead to increased temperature in the core of London. Solar panels could potentially provide sufficient energy for running air conditioning all over London, creating comfortable indoor environments, and green roofs could reduce temperatures during the day. We argue that such inter-comparisons should guide future decision making.

1 Introduction

Cities have become a focus of attention for climate adaptation and mitigation solutions in the face of the increasing challenges induced by global urbanization and climate warming. More specifically, interest in urban temperatures has been growing, as cities are known to alter the local climate to which urban citizens are exposed (e.g., through the Urban Heat Island (UHI) effect (Oke et al., 2017) and thereby impact their thermal comfort and climate-related mortality (Arbutnott et al., 2016; Gasparrini et al., 2015; Masselot et al., 2023). During hot spells, higher mortality and increased likelihood of thermal discomfort are generally observed, hence calling for adaptative and mitigative measures to reduce outdoor air temperatures at the city scale (Heaviside et al., 2016; Salmond et al., 2016; Jungman et al., 2023).

To respond to these risks to public health, passive and active strategies have been proposed to cool down the urban indoor and outdoor environments, and to provide indoor sheltering for the most vulnerable. Commonly proposed urban passive cooling strategies, which should be considered as measures that directly lower the temperatures, include increased urban vegetation, roofs incorporating vegetation (known as green roofs), or the deployment of highly reflective roofs, known as cool roofs (Santamouris et al., 2017). Changes to roofs can reduce indoor temperatures or cooling needs in a building, but when deployed at scale can also reduce the outdoor air temperature and associated heat related mortality (Virk et al., 2014; H. Macintyre & Heaviside, 2019). Concerning active strategies, meaning measures that transform the incoming radiation, into energy for ex-

ample, or that mechanically change the temperatures, air conditioning (AC) is one of the most effective for protection against heat, although it can require large amounts of electrical power and increase outdoor temperatures (Salamanca et al., 2014; Stone Jr et al., 2021). Where passive strategies are unviable or insufficient, AC may be the only option, but it remains the primary solution in some high-income countries despite its problems, like in Phoenix, Arizona (Stone Jr et al., 2021). Lastly, while rooftop solar photovoltaic (PV) panels are primarily considered as a source of electrical power, they can also be considered as a passive-active strategy for impacting outdoor and indoor air temperature by increasing the roofs' albedo and by transforming incoming solar radiation into electrical power that can be used to run the AC system (Salamanca et al., 2016; Ma et al., 2017). Debate on their effectiveness as cooling strategies is currently ongoing (Sailor et al., 2021) and calls for more local studies to understand their costs and benefits at different latitudes.

In general, the impact of these strategies has been quantified for a variety of cities across the world (e.g., Chicago (Sharma et al., 2016), Madrid (Salamanca et al., 2014) or Singapore (Yang & Bou-Zeid, 2019)) and physical mechanisms that lead to the cooling/heating of the cities are well understood (Nazarian et al., 2022). Nonetheless, city-specific quantification of these impacts should still be performed to inform the general audience. In addition, few studies have explored in detail how these interventions impact the surface energy balance (SEB) and how these alterations at the city scale can relate to local changes in air temperatures (Broadbent et al., 2020; Krayenhoff et al., 2021). Furthermore, impact studies of urban scale interventions are often challenged by the lack of geographical data on realistic extent at which each intervention could be applied. This means that it is often impossible to give a realistic estimate of the potential impact that interventions can have at a city-scale.

Large-scale controlled trials of urban interventions are, so far, rare, and much of the evidence for their effectiveness comes from modelling. Urban climate models (UCMs) embedded in regional climate models (RCMs), after proper model evaluation, can provide an accurate estimation of the impact on the of urban areas on local climate, as well as the impact of interventions (Krayenhoff et al., 2021). Indeed, most UCMs are coupled to a variety of intervention strategies (e.g., green roofs and solar panels (Zonato et al., 2021) or AC systems (Salamanca et al., 2010; Salamanca & Martilli, 2010)).

In this study, we use detailed urban climate modelling, as well as recent data on the actual coverage of passive(-active) interventions, to estimate the impact on outdoor temperature of several interventions during hot summer days within the Greater London Authority boundaries. By considering a set of scenarios that include both active and passive heat adaptation strategies, our study makes a comprehensive analysis of the impact of each strategy on the outdoor air temperature and the SEB of the Greater London area. Because we expect some of these strategies to not be applicable to all buildings, like green roofs or solar PV panels, we provide realistic scenarios to compare the hypothetical maximum with realistic implementation. We also discuss the potential energy costs and benefits of air conditioning and PV, and the advantages and disadvantages of vegetation-based strategies.

2 Methods

We ran a set of regional climate simulations with a complex three-dimensional urban scheme activated on the two hottest days of the summer of 2018. Our model is based on an existing set up, run over the whole summer of 2018 (Brousse et al., 2023) and is evaluated against a set of in situ observations of air temperature. Although our model is run over the whole of south-east England to accurately represent the interaction between local interventions and mesoscale atmospheric circulations, we focus our study on

118 the Greater London Authority (GLA) metropolitan area only, because our realistic in-
119 terventions are valid for that specific area.

120 2.1 Model configuration

121 We focus our study on the two hottest days of summer 2018, the 26th and the 27th
122 of July, as 2018 was the hottest summer on average for the United Kingdom (McCarthy
123 et al., 2019). Our RCM is run for 3 consecutive days, starting on the 25th of July 2018,
124 and considers that day as spin-up time. We use the Weather Research Forecast (WRF) v4.3
125 RCM to physically represent the potential impact of urban temperature reduction in-
126 terventions at a high spatio-temporal resolution of 1 km with hourly outputs over south-
127 east England; following the previous model domain of Brousse et al. (2023). The impact
128 of cities on the local climate is simulated by activating the Building Energy Parameter-
129 ization (BEP; Martilli et al. (2002)) model with its coupled Building Energy Model (BEM;
130 Salamanca et al. (2010); Salamanca and Martilli (2010)), hereafter referred as BEP-BEM.

131 We consider BEP-BEM to be optimal for our study as it is one of the most refined
132 models that permits the modelling of urban areas at meso-scale. It is a complex 3-dimensional
133 UCM that divides the urban environment and atmosphere into several key components
134 and layers. For instance, BEP-BEM allows the users to define building characteristics
135 at roof and wall levels and street characteristics separately and represents the building
136 impact on the urban boundary layer with a subgrid vertical resolution of 5 m. This is
137 particularly important for evaluation of rooftop interventions as Zonato et al. (2021) showed
138 that the impact of interventions is dependent on building height. Using BEP-BEM thereby
139 avoids over-simplification of the changes in urban properties induced by the interven-
140 tion – something that would happen in a bulk-level UCM, for example. It also subdivi-
141 des the RCM grid into an urban and a natural tile where surface fluxes are calculated
142 separately before being merged, hence allowing for an appropriate estimation of the im-
143 pact of changes in vegetation cover. In our control simulation, the natural tile is han-
144 dled by the Noah-MP land surface model (Niu et al., 2011) and is considered as a mix-
145 ture of cropland and grassland. Lastly, BEP-BEM offers a set of dynamic models for ac-
146 tive and passive strategies for cooling outdoor and indoor environments, including green
147 roofs, solar PV panels, and air conditioning (Salamanca et al., 2011; Zonato et al., 2021).

148 We use the same model dynamics and physics set up as (Brousse et al., 2023) as
149 these have shown good agreement with observations for the whole summer 2018. For the
150 urban canopy parameters and the land cover classification we use the defaults MODIS
151 land cover from WRF and categorize the urban pixels to Local Climate Zones (LCZ; Stewart
152 and Oke (2012)) following the WUDAPT-TO-WRF strategy (Brousse et al., 2016; De-
153 muzere et al., 2022) to provide the WRF model with spatially explicit urban morpho-
154 logical parameters. The building characteristics are provided per LCZ and are the same
155 than in Brousse et al. (2023).

156 2.2 Model evaluation

157 To evaluate the model outputs in parts of the urban environment deprived of of-
158 ficial weather data sources, we follow the same evaluation strategy as in Brousse et al.
159 (2023) by taking advantage of the dense network of personal weather stations (PWS)
160 in the city of London (Brousse et al., 2022). Our model is thereby evaluated against of-
161 ficial automatic weather stations (AWS) from the MIDAS Met Office network (Sunter,
162 2021; UKMO, 2021) and quality-checked crowd-sourced PWS from the Netatmo com-
163 pany (<https://netatmo.org>). More details on the data acquisition and on the ration-
164 ale defending the use of privately owned PWS for UCM evaluation can be found in Brousse
165 et al. (2023). All our crowd-sourced weather data has been quality-checked and filtered
166 using a common statistical method provided by the *Crowd-QC* package from Grassmann
167 et al. (2018); also see Napoly et al. (2018).

168 We evaluate our model with several statistical indicators including the squared Pear-
 169 son’s coefficient of correlation (r^2), Spearman’s coefficient of correlation, root mean squared
 170 error (RMSE), mean absolute error (MAE), and mean bias (MB). The evaluation is car-
 171 ried out over the whole domain to ensure that the model is performing well in all urban
 172 settings and to ensure that our choice for the GLA area is not biased towards higher model
 173 performances.

174 2.3 Temperature interventions

175 We define a set of 9 passive interventions that are theoretically applicable over the
 176 Greater London metropolitan area. Most of our interventions consist of roof interven-
 177 tions and each intervention is compared to a control run which consists of the same model
 178 set up as in Brousse et al. (2023), and which did not consider any heat intervention strat-
 179 egy. Therefore, the control run is simply representative of the impact of urban areas on
 180 the local climate while all subsequent scenarios estimate the impact of each intervention
 181 on that urban climate. We also simulate the effect of widespread use of active cooling
 182 in the form of air conditioning.

183 Our scenarios consist of: 1) the deployment of cool roofs by increasing the albedo
 184 of each urban class to 0.85 (CR_{100}); 2) the deployment of green roofs composed of se-
 185 dum vegetation in all urban classes by activating the green roof model from Zonato et
 186 al. (2021) with its default irrigation hours from 23.00 UTC to 01.00 UTC (GR_{100}); 3)
 187 the deployment of solar PV panels on all urban roofs by activating the solar PV model
 188 from (Zonato et al., 2021) in its default version (PV_{100}); 4) the installation of air con-
 189 ditioning systems in all buildings and at all floors by activating the air conditioning model
 190 embedded in the BEM model in its default version (Salamanca et al. (2010); Salamanca
 191 and Martilli (2010); AC_{100}); and 5) the change of the urban vegetation from cropland
 192 and grass land to urban deciduous trees as implemented in NOAH-MP (Niu et al. (2011),
 193 VG_{100}). These scenarios aim to represent the maximum impact that one intervention
 194 can have on the city’s temperatures and SEB. More information about each model can
 195 be found in the appendix Appendix A.

196 We also test the impact of certain scenarios that are based on more achievable or
 197 tailored interventions. These scenarios are: 1) the potential for solar PV deployment over
 198 roofs of existing buildings in the Greater London taking into account roof slope and over-
 199 shadowing derived from the London solar opportunity map (PV_{pot} ; Steadman et al. (2020));
 200 2) the potential for retrofit of green roofs on existing buildings in Greater London ob-
 201 tained through a detailed geographical information system analysis (C. Simpson et al.
 202 (2022); GR_{pot}); 3) the installation of cool roofs in compact and open mid-rise areas (Lo-
 203 cal Climate Zones 2 and 5) from the centre of London (CR_{mid}); 4) in open low rise ar-
 204 eas (Local Climate Zone 6; CR_{low}) from the residential suburbs of London; and 5) in com-
 205 mercial and industrial areas of London (Local Climate Zone 8; CR_{cid}).

206 Potential conjoint use of green and cool roofs with solar panels have been proposed
 207 to increase solar panel efficiency while reducing urban surface temperatures in particu-
 208 lar (Fleck et al., 2022; Schindler et al., 2018), but it was not possible to analyse this
 209 in our model. Sedum was modelled for the green roofs due to its increasing presence on
 210 new buildings in London (C. Simpson et al., 2022), but other plants such as turf may
 211 have a different impact.

212 To integrate the spatially-explicit variation in solar PV panels and green roofs cov-
 213 erage from the scenarios PV_{pot} and GR_{pot} , we adapted the code from Zonato et al. (2021)
 214 and added a field to the WRF geographical input files. These two scenarios also explain
 215 why our study focuses only on GLA area as these are bounded to its administrative bound-
 216 aries and are not applicable to the whole south-east of England which composes our WRF
 217 domain. Lastly, we expect the impacts to be more prominent in London due to its higher
 218 urban heat island intensity (Bassett et al., 2021; Brousse et al., 2022).

3 Results

During the two hottest summer days of 2018, our model configuration presents a warm bias estimated at $0.72\text{ }^{\circ}\text{C}$ on average in urban PWS and of $1.76\text{ }^{\circ}\text{C}$ amongst official AWS (Fig. B1). Such a difference may be explained by the fact that official AWS are often located in open fields like parks or airports, which further supports the use of PWS for model evaluation (Hammerberg et al., 2018; Brousse et al., 2023). Metrics estimating the model error are better on average across urban PWS compared to AWS, with average values of $2.39\text{ }^{\circ}\text{C}$ compared to $2.84\text{ }^{\circ}\text{C}$ for RMSE, and of $1.94\text{ }^{\circ}\text{C}$ compared to $2.33\text{ }^{\circ}\text{C}$ for MAE. The model seems however to be better correlated temporally at AWS locations with coefficients of 0.91 against 0.88 for Spearman’s r and of 0.84 against 0.81 for Pearson’s r^2 . We consider our model performance to be acceptable, especially considering the high level of confidence in the temporal evolution of the air temperature at 2 m (T_2) which is directly correlated to the SEB. We thereby assume that our model is trustworthy for each SEB component as we cannot evaluate them without access to flux tower measurements.

In general, we find that cool roofs are the more effective interventions to reduce the heat in the GLA boundaries with a 2-days average cooling of $\sim 1.2\text{ }^{\circ}\text{C}$ going up to $\sim 2.0\text{ }^{\circ}\text{C}$ in certain locations (Fig. 1 and 2). All cool roofs interventions contribute to a reduction of T_2 on average (Fig. 2) with major impacts found in the south and the east of GLA if deployed at 100 % (CR_{100} ; Fig. 1). Horizontal advection of cooling depends on the location where cool roofs are implemented. For instance, if cool roofs are implemented in more central and mid-rise areas of London, more open and low-rise residential areas appear to benefit from this cooling by $\sim 0.15\text{ }^{\circ}\text{C}$ on average. Interestingly, implementing cool roofs on residential low-rise buildings only (CR_{low}) does not cool the central parts of the cities despite providing a substantial outdoor T_2 reduction by up to $\sim 2\text{ }^{\circ}\text{C}$ in the south-east. Establishing cool roofs only on industrial and commercial areas (CR_{cid}) only results in a marginal heat reduction on average over London but does lead to a heat reduction of up to $\sim 1\text{ }^{\circ}\text{C}$ in large industrial areas of east-London. Compared to cool roofs, we find more minor decreases in T_2 of $\sim 0.3\text{ }^{\circ}\text{C}$ on average related to both the change in natural cover from a mixture of croplands and grasslands to deciduous trees (VG_{100}) and to the installation of solar panels over the whole city (PV_{100}). Stronger cooling effects are found in the east of London for these two scenarios. Lastly, despite a high spatial heterogeneity of the cooling provided by the remaining interventions (green roofs (GR_{100} and GR_{pot}); realistic retrofit potential of solar panels (PV_{pot}), none is found to be beneficial for T_2 reduction on average for the whole metropolitan area (Fig. 2). Using air conditioning for maintaining the indoor temperature of all London’s building stock to $21\text{ }^{\circ}\text{C}$ (AC_{100}) induces a heating of central London by up to $\sim 1\text{ }^{\circ}\text{C}$ and of $\sim 0.15\text{ }^{\circ}\text{C}$ on average for the whole city.

By relating the change in T_2 of all urban pixels in the GLA area to their relative SEB components we find that the large decrease in T_2 induced by the full scale deployment of cool roofs (CR_{100}) is linked to an increase in albedo by ~ 0.3 and a reduction of sensible heat fluxes emitted by the city of $\sim 30\text{ }W\cdot m^{-2}$, on average. Notably, however, we observe an increase in incoming solar radiation when deploying cool roofs over the whole city (CR_{100}) and in open residential areas (CR_{low}) as well as when implementing solar panels over the whole city (PV_{100}); something that was also recently observed by Valencia et al. (2023). This could be related to a lower convection due to the cooler surfaces, although major changes in the average cloud cover in the whole atmospheric column are not perceived. Higher emissions of sensible heat fluxes by $\sim 5\text{ }W\cdot m^{-2}$ are found in both the air conditioning (AC_{100}) and the deciduous tree (VG_{100}) scenarios. While the former is expected as AC systems remove the accumulated indoor heat outside mostly in the form of sensible heat, the latter is counter intuitive as more vegetated areas are expected to mostly transform energy in the form of latent heat fluxes via evapotranspiration – something that is modelled with an average increase of latent heat fluxes

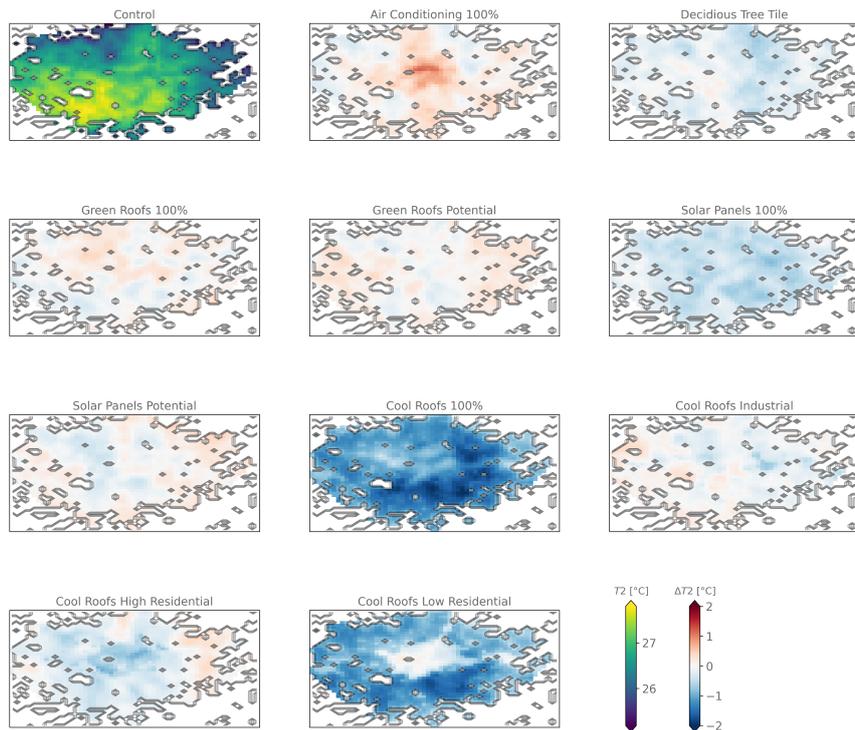


Figure 1. Average 2 m air temperature of the control run (T_2 ; top left panel) and the average difference of each intervention (ΔT_2). Grey contours are WRF urban pixels (classified as an urban LCZ).

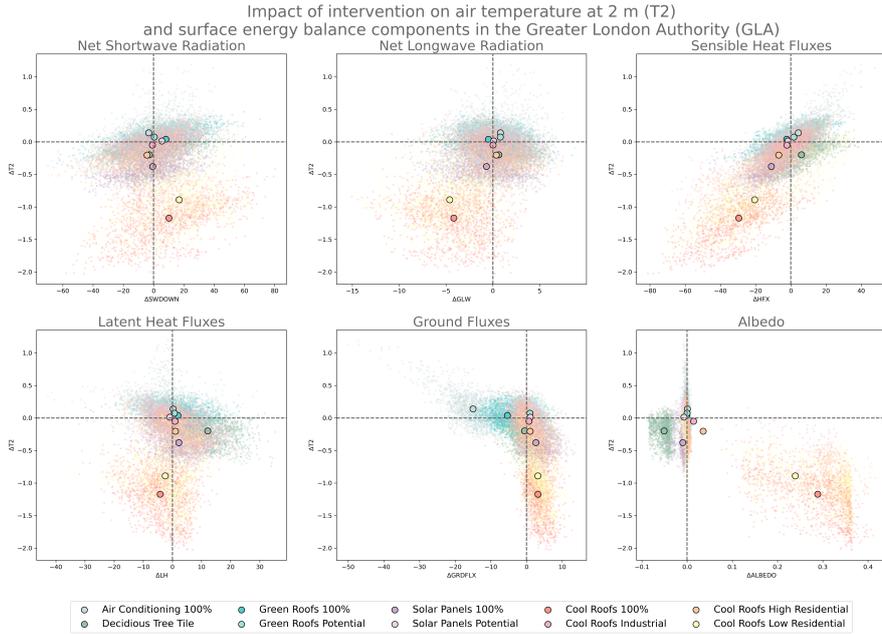


Figure 2. Time-averaged impact of each intervention on each surface energy balance component and the respective change in air temperature at 2 m for all urban pixels (small dots). Average impact is represented by the large dots.

272 by $\sim 12 \text{ W}\cdot\text{m}^{-2}$. We however observe a decrease in surface albedo by 0.05 on average
 273 in parallel which could explain this phenomenon.

274 Despite an increase of T_2 induced by green roofs of less than $0.1 \text{ }^\circ\text{C}$ on average,
 275 hourly spatial averages of changes in T_2 show that if green roofs were implemented across
 276 the whole city (GR_{100}), temperatures could be decreased by ~ 0.5 to $\sim 0.8 \text{ }^\circ\text{C}$ during hot
 277 hours of the day (11.00 UTC to 14.00 UTC; Fig. 3 and Fig. C1 to C2). Nonetheless, tem-
 278 peratures would also be increased at times of daily minimum T_2 by $\sim 0.5 \text{ }^\circ\text{C}$ (03.00 UTC
 279 to 06.00 UTC, Fig. 3). This is consistent with findings by Zonato et al. (2021) and ex-
 280 plained by an increased release of latent fluxes during the day but a higher storage of
 281 heat fluxes at night due to green roofs' thermal mass. The current potential for green
 282 roofs in London (GR_{pot}) would not reduce the temperature by more than $0.1 \text{ }^\circ\text{C}$ on av-
 283 erage across the period. During the hotter hours of the day, the impact of cool roofs (CR_{100})
 284 can be 2.5 times and the one from cool roofs over open-low rises (CR_{low}) 2.0 times greater
 285 than the one of green roofs. Both these scenarios (CR_{100} and CR_{low}) also contribute to
 286 a constant reduction of temperature of at least $0.5 \text{ }^\circ\text{C}$ throughout the day. This is cor-
 287 related to a high reduction of emitted sensible heat fluxes and of the positive intakes of
 288 ground fluxes throughout daytime hours (Fig. C1 to C2). Notably, the temperature re-
 289 duction across the whole city induced by cool roofs over central mid-rises from London
 290 (CR_{mid}) reaches a peak at later hours, around 15.00 UTC, a few hours after the peak in
 291 reduction of sensible heat fluxes reached at noon. Higher temperature reduction by
 292 $\sim 0.8 \text{ }^\circ\text{C}$ is observed in the evening if vegetation is changed to deciduous trees (VG_{100}).
 293 During the day, no tangible impact is found in this case, as half of the urban pixels ob-
 294 serve an increase in temperature while the other half observes a decrease. This may be
 295 because higher levels of sensible heat fluxes are emitted; more latent heat fluxes are also
 296 emitted throughout the day. Both full scale and realistic solar panel implementation (PV_{100}

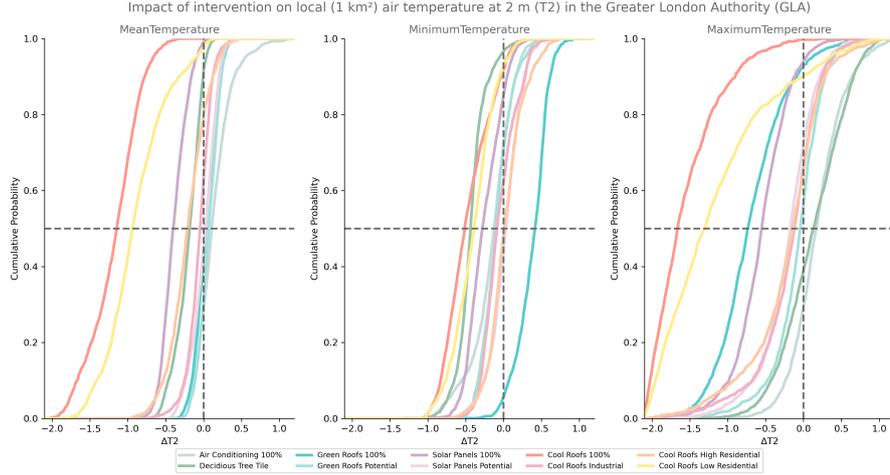


Figure 3. Cumulative distributions of change in air temperature at 2 m for different intervention scenarios. Distributions are of change relative to the reference scenario in average daily mean, daily minimum, and daily maximum 2 m air temperature across grid cells. Distributions are from 200 bin histograms, normalised to unity and based on modelled air temperature on 26th and the 27th of July 2018 in London. The horizontal dashed line indicates the median change. The vertical dashed line indicates zero change.

297 and PV_{pot}) behave similarly to cool roofs by decreasing the emitted sensible heat fluxes
 298 and reducing the intake of fluxes to the ground during daytime hours. The temperature
 299 reduction of up to ~ 1.0 °C, for the full-scale deployment, however, only happens dur-
 300 ing the late afternoon and the evening. At time of daily maximum temperatures, PV_{100}
 301 reduces the temperature in similar ways to GR_{100} (3 and Fig. C1 to C2).

302 4 Discussion

303 Using the WRF BEP-BEM mesoscale urban climate over the Greater London metropoli-
 304 tan area during the two hottest days of 2018, we performed one of the first comprehen-
 305 sive evaluations of the impact of building and street level interventions tailored at re-
 306 ducing outdoor and indoor heat on the outdoor temperature and the surface energy bal-
 307 ance in London.

308 We find that interventions such as cool roofs, green roofs, rooftop solar PV, and
 309 changing natural spaces to deciduous woodland can contribute to a cooling of outdoor
 310 temperatures at hyper-local or city scales (also see Fig. C3 to C8, while widespread use
 311 of air conditioning could lead to substantial increases in outdoor temperature in the cen-
 312 tre of London. On average, cool roofs would be the most effective way of reducing out-
 313 door temperature at the city scale, in line with other modelling studies (Sharma et al.,
 314 2016; Yang & Bou-Zeid, 2019; Tan et al., 2023). At the maximum, this temperature re-
 315 duction is between 3.2 °C and 2.8 °C at times when daily spatial average temperatures
 316 reached 33 °C and 37 °C, respectively (Fig. C2), similar to the cooling found in Phoenix
 317 and New York (Sinsel et al., 2021). In comparison, if green roofs or rooftop solar PV pan-
 318 els were deployed over all buildings in the Greater London area, the temperature reduc-
 319 tion would have never reached more than ~ 1.0 °C for both days; considering that overnight,
 320 green roofs would increase the average temperature due to their increased heat-storage
 321 capacity that increases heat release at night (Tan et al., 2023). Of note, our green roofs

322 are modelled with sedum vegetation as they are the most common in London; other de-
323 signs may lead to other effects as Zonato et al. (2021) found an increased reduction us-
324 ing grass-based green roofs. Conversely, we find that changing the modelled vegetation
325 type of urban natural spaces from grasslands/croplands to deciduous trees would reduce
326 outdoor temperatures only during the night and found mixed effects during the day, pos-
327 sibly due to an increase in the surface albedo which leads to an increased sensible heat
328 emission during the day. Latent heat would also increase as expected from more evap-
329 otranspiration and would be equivalent to the increase in latent heat from green roofs
330 deployed at full scale. Although cooling via evapotranspiration reduces air temperature,
331 it increases water vapor in the atmosphere; this could be beneficial for avoiding water
332 stress of vegetation but affect the heat stress from city's inhabitants and should be eval-
333 uated using a set of multiple heat stress metrics (C. H. Simpson et al., 2023).

334 City scale impacts of any intervention are reduced when applied to a smaller area
335 or fewer buildings. For cool roofs, we find that although major impacts are found when
336 applied over low-rises residential areas of London, already applying them over central
337 mid-rises of the city could reduce the temperature in surrounding residential areas. Be-
338 cause of the relatively small area covered by industrial areas in London, applying cool
339 roofs over them would only marginally contribute to a heat reduction of the city, con-
340 tradictory to results in Birmingham where commercial and industrial areas cover an im-
341 portant part of the city centre (H. Macintyre & Heaviside, 2019). They could also be con-
342 sidered less effective as industrial buildings already observe higher albedo than more res-
343 idential areas (Brousse et al., 2023). Cool roofs may act to reduce indoor temperatures
344 in summer, but also potentially in winter, thus increasing winter heating demand. There-
345 fore, it is important that they are installed in conjunction with insulation in homes (Taylor
346 et al., 2018). However, in winter, cool roofs have less of a cooling effect outdoor temper-
347 atures than they do in summer (H. L. Macintyre et al., 2021). While we find that full
348 application of rooftop solar PV leads to some cooling on average, realistic application,
349 constrained by existing building characteristics, is not enough to have an appreciable ef-
350 fect on the temperature. Similarly, full application of sedum green roofs can reduce day-
351 time temperatures, but the potential for retrofit of existing buildings with sedum green
352 roofs is not enough to produce a substantial cooling effect at the city scale.

353 Although air conditioning would increase the outdoor temperature, it could still
354 protect vulnerable populations from overheating indoors. Through our modelling exper-
355 iment, we find that if solar panels were implemented in a realistic manner on the rooftops
356 of existing buildings, the energy production, peaking at ~ 22.5 MW at time of higher solar
357 incidence (11.00 UTC) and with an average production of ~ 7 MWh, would cover the
358 energetic consumption induced by AC systems which never goes above ~ 9 MW during
359 the hottest hours (15.00 UTC to 17.00 UTC), averaging to ~ 4.5 MWh. Under the as-
360 sumption that deployment of solar panels is not constrained by existing buildings, elec-
361 tric power production is 75 % greater than the realistic retrofit scenario, and outdoor
362 average temperature is reduced by ~ 0.3 °C while temperature reduction is negligible in
363 the realistic case. This temperature reduction should be carefully interpreted as several
364 studies have found that solar panels could heat up the surrounding environment (Sailor
365 et al., 2021). Nonetheless, these local impacts may be dependent on many factors, in-
366 cluding the building height and the building plan area fraction (Zonato et al., 2021), or
367 the increase in rooftop albedo. In our case, we argue that the capture of solar radiation
368 by the solar PVs is sufficiently important to lower the expected intake of radiation by
369 the building and that most of it is transformed in the form of electric energy rather than
370 in sensible heat over the roof. In the evening and at night, the low thermal capacity of
371 solar PVs explains the higher cooling capacity of the urban environment. In any case,
372 our results call for more investigation on this specific topic as other studies, like (Tan
373 et al., 2023), found a small night-time increase in temperature over Chicago when im-
374 plementing solar PV over buildings' roofs. Hence mechanisms explaining solar PV ef-
375 ficiency in our model need to be further explored.

376 Several limitations apply to this study. First, we used only one UCM, and another
377 UCM may produce different results depending on its dynamics and physics (Krayenhoff
378 et al., 2021). Second, we focused our study only over two hot summer days due to com-
379 putational limitations. In general, performing such comprehensive analysis with multi-
380 ple scenarios comes at high computational costs, which often prevents the generalization
381 of the results to expected average seasonal impacts of these interventions or to other cities
382 at similar latitudes. Third, despite the fact that the 26th and the 27th of July were ex-
383 tremely hot days during the summer 2018 in the United Kingdom, full clear-sky condi-
384 tion is absent from our model simulations, which could, for instance, partially explain
385 why solar PV do not lead to a heating of the city during the day. Some of the results
386 observed in this study concerning the surface energy balance may therefore simply be
387 related to the overpass of clouds; although we did not find a particular difference in cloud
388 coverage between our control simulation and our sensitivity tests. Fourth, as discussed
389 above, our study only focuses on outdoor temperatures and does not consider the ben-
390 efits or the detriments caused by these interventions on other aspects (e.g., biodiversity,
391 indoor temperature, increase in relative humidity, wind circulation, etc.). Lastly, all our
392 estimated impacts should be interpreted cautiously as the average temperature differ-
393 ences resulting from the interventions are smaller than the model MAE with respect to
394 observations (Krayenhoff et al., 2021). Despite these limitations, we expect our results
395 to be at the least indicative of which interventions are capable of producing the great-
396 est cooling effect in London.

397 Notwithstanding, the outcomes of this study also have to be put in perspective against
398 the costs related to the deployment of each of these interventions which should be weighted
399 according to a cost-benefit analysis.

400 5 Conclusions

401 Adaptation to urban heat is of increasing priority. Our modelling study suggests
402 that widespread adoption of cool roofs could be efficient at reducing urban temperatures,
403 and therefore adverse health impacts, during hot spells. Rooftop solar PV may also pro-
404 vide a small amount of cooling as an additional benefit to power production; green roofs
405 appear less effective in reducing temperatures but have other environmental benefits. Fur-
406 thermore, relying on AC will lead to increased outdoor temperatures, despite improved
407 thermal comfort indoors. It is reasonable to think that there will be a mixture of adap-
408 tation measures implemented across the city of London, and their applicability and im-
409 pact should be further investigated. Our research is a first step towards this and will be
410 of interest to city planners and decision-makers within the built environment sector.

411 6 Open Research

412 The simulations done in this research were performed using the WRF model v4.4
413 (<https://github.com/wrf-model/WRF.git>). The related outputs presented in this re-
414 search and codes used to plot them are available following this DOI: 10.5281/zenodo.8333363

415 Acknowledgments

416 CH is supported by a NERC fellowship (NE/R01440X/1) and acknowledges funding for
417 the HEROIC project (216035/Z/19/Z) from the Wellcome Trust, which funds OB and
418 CS.

419 OB designed the study and led the conception of the manuscript with the support
420 of CH and CS. OB was responsible for the WRF modelling, the model evaluation and
421 the analysis. CS provided support in the integration of green roofs and solar PV inputs
422 into WRF. AZ implemented a spatially-explicit input of solar PV and green roofs cov-
423 erages in WRF. AZ and AM offered guidance in the set-up of the WRF model v4.4 and

424 in the analysis of the outputs. JT and MD provided support in the design of the study
425 and in the acquisition of knowledge about building interventions. All authors contributed
426 to the writing of the manuscript.

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1 **Cool roofs could be most effective at reducing outdoor**
2 **urban temperatures in London compared with other**
3 **roof top and vegetation interventions: a mesoscale**
4 **urban climate modelling study**

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11 **Key Points:**

- 12 • City scale deployment of cool roofs have the highest impact on 2 m air temper-
13 ature
14 • Green roofs do not decrease daily average temperature but have a daytime cool-
15 ing effect
16 • Solar PV can reduce temperatures in London by capturing sensible heat, gener-
17 ating electrical power and reducing sensible heat flux

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Abstract

Comprehensive studies comparing impacts of building and street level interventions on air temperature at metropolitan scales are still lacking despite increased urban heat-related mortality and morbidity. We therefore model the impact of 9 interventions on air temperatures at 2 m during 2 hot days from the summer 2018 in the Greater London Authority area using the WRF BEP-BEM climate model. We find that on average cool roofs most effectively reduce temperatures (~ -1.2 °C), outperforming green roofs (~ 0 °C), solar panels (~ -0.3 °C) and street level vegetation (~ -0.3 °C). Application of air conditioning across London increase air temperatures by $\sim +0.15$ °C but related energetic consumption could be covered by energy production from solar panels. Current realistic deployments of green roofs and solar panels are ineffective at large scale reduction of temperatures. We provide a detailed decomposition of the surface energy balance to explain changes in air temperature and guide future decision-making.

Plain Language Summary

Multiple common city scale passive and active interventions exist to reduce urban population's exposure to extreme heat during hot spells. Nonetheless, a proper comparison of the effect that each of these interventions may have on the temperatures experienced within large cities is missing. Additionally, energetic mechanisms that lead to outdoor temperature changes are often not detailed and could lead to detrimental effects for local populations, such as indirect increase of relative humidity or outdoor temperatures. Our study, focusing over London, compares several common interventions through a modelling experiment and finds that cool roofs largely outperform other interventions during the two hottest days of the summer 2018. We also find that green roofs are ineffective on average and that solar panels and tree vegetation would only marginally change temperature exposures. Large scale deployment of air conditioning would lead to increased temperature in the core of London. Solar panels could potentially provide sufficient energy for running air conditioning all over London, creating comfortable indoor environments, and green roofs could reduce temperatures during the day. We argue that such inter-comparisons should guide future decision making.

1 Introduction

Cities have become a focus of attention for climate adaptation and mitigation solutions in the face of the increasing challenges induced by global urbanization and climate warming. More specifically, interest in urban temperatures has been growing, as cities are known to alter the local climate to which urban citizens are exposed (e.g., through the Urban Heat Island (UHI) effect (Oke et al., 2017) and thereby impact their thermal comfort and climate-related mortality (Arbutnott et al., 2016; Gasparrini et al., 2015; Masselot et al., 2023). During hot spells, higher mortality and increased likelihood of thermal discomfort are generally observed, hence calling for adaptative and mitigative measures to reduce outdoor air temperatures at the city scale (Heaviside et al., 2016; Salmond et al., 2016; Jungman et al., 2023).

To respond to these risks to public health, passive and active strategies have been proposed to cool down the urban indoor and outdoor environments, and to provide indoor sheltering for the most vulnerable. Commonly proposed urban passive cooling strategies, which should be considered as measures that directly lower the temperatures, include increased urban vegetation, roofs incorporating vegetation (known as green roofs), or the deployment of highly reflective roofs, known as cool roofs (Santamouris et al., 2017). Changes to roofs can reduce indoor temperatures or cooling needs in a building, but when deployed at scale can also reduce the outdoor air temperature and associated heat related mortality (Virk et al., 2014; H. Macintyre & Heaviside, 2019). Concerning active strategies, meaning measures that transform the incoming radiation, into energy for ex-

ample, or that mechanically change the temperatures, air conditioning (AC) is one of the most effective for protection against heat, although it can require large amounts of electrical power and increase outdoor temperatures (Salamanca et al., 2014; Stone Jr et al., 2021). Where passive strategies are unviable or insufficient, AC may be the only option, but it remains the primary solution in some high-income countries despite its problems, like in Phoenix, Arizona (Stone Jr et al., 2021). Lastly, while rooftop solar photovoltaic (PV) panels are primarily considered as a source of electrical power, they can also be considered as a passive-active strategy for impacting outdoor and indoor air temperature by increasing the roofs' albedo and by transforming incoming solar radiation into electrical power that can be used to run the AC system (Salamanca et al., 2016; Ma et al., 2017). Debate on their effectiveness as cooling strategies is currently ongoing (Sailor et al., 2021) and calls for more local studies to understand their costs and benefits at different latitudes.

In general, the impact of these strategies has been quantified for a variety of cities across the world (e.g., Chicago (Sharma et al., 2016), Madrid (Salamanca et al., 2014) or Singapore (Yang & Bou-Zeid, 2019)) and physical mechanisms that lead to the cooling/heating of the cities are well understood (Nazarian et al., 2022). Nonetheless, city-specific quantification of these impacts should still be performed to inform the general audience. In addition, few studies have explored in detail how these interventions impact the surface energy balance (SEB) and how these alterations at the city scale can relate to local changes in air temperatures (Broadbent et al., 2020; Krayenhoff et al., 2021). Furthermore, impact studies of urban scale interventions are often challenged by the lack of geographical data on realistic extent at which each intervention could be applied. This means that it is often impossible to give a realistic estimate of the potential impact that interventions can have at a city-scale.

Large-scale controlled trials of urban interventions are, so far, rare, and much of the evidence for their effectiveness comes from modelling. Urban climate models (UCMs) embedded in regional climate models (RCMs), after proper model evaluation, can provide an accurate estimation of the impact on the of urban areas on local climate, as well as the impact of interventions (Krayenhoff et al., 2021). Indeed, most UCMs are coupled to a variety of intervention strategies (e.g., green roofs and solar panels (Zonato et al., 2021) or AC systems (Salamanca et al., 2010; Salamanca & Martilli, 2010)).

In this study, we use detailed urban climate modelling, as well as recent data on the actual coverage of passive(-active) interventions, to estimate the impact on outdoor temperature of several interventions during hot summer days within the Greater London Authority boundaries. By considering a set of scenarios that include both active and passive heat adaptation strategies, our study makes a comprehensive analysis of the impact of each strategy on the outdoor air temperature and the SEB of the Greater London area. Because we expect some of these strategies to not be applicable to all buildings, like green roofs or solar PV panels, we provide realistic scenarios to compare the hypothetical maximum with realistic implementation. We also discuss the potential energy costs and benefits of air conditioning and PV, and the advantages and disadvantages of vegetation-based strategies.

2 Methods

We ran a set of regional climate simulations with a complex three-dimensional urban scheme activated on the two hottest days of the summer of 2018. Our model is based on an existing set up, run over the whole summer of 2018 (Brousse et al., 2023) and is evaluated against a set of in situ observations of air temperature. Although our model is run over the whole of south-east England to accurately represent the interaction between local interventions and mesoscale atmospheric circulations, we focus our study on

118 the Greater London Authority (GLA) metropolitan area only, because our realistic in-
119 terventions are valid for that specific area.

120 2.1 Model configuration

121 We focus our study on the two hottest days of summer 2018, the 26th and the 27th
122 of July, as 2018 was the hottest summer on average for the United Kingdom (McCarthy
123 et al., 2019). Our RCM is run for 3 consecutive days, starting on the 25th of July 2018,
124 and considers that day as spin-up time. We use the Weather Research Forecast (WRF) v4.3
125 RCM to physically represent the potential impact of urban temperature reduction in-
126 terventions at a high spatio-temporal resolution of 1 km with hourly outputs over south-
127 east England; following the previous model domain of Brousse et al. (2023). The impact
128 of cities on the local climate is simulated by activating the Building Energy Parameter-
129 ization (BEP; Martilli et al. (2002)) model with its coupled Building Energy Model (BEM;
130 Salamanca et al. (2010); Salamanca and Martilli (2010)), hereafter referred as BEP-BEM.

131 We consider BEP-BEM to be optimal for our study as it is one of the most refined
132 models that permits the modelling of urban areas at meso-scale. It is a complex 3-dimensional
133 UCM that divides the urban environment and atmosphere into several key components
134 and layers. For instance, BEP-BEM allows the users to define building characteristics
135 at roof and wall levels and street characteristics separately and represents the building
136 impact on the urban boundary layer with a subgrid vertical resolution of 5 m. This is
137 particularly important for evaluation of rooftop interventions as Zonato et al. (2021) showed
138 that the impact of interventions is dependent on building height. Using BEP-BEM thereby
139 avoids over-simplification of the changes in urban properties induced by the interven-
140 tion – something that would happen in a bulk-level UCM, for example. It also subdivi-
141 des the RCM grid into an urban and a natural tile where surface fluxes are calculated
142 separately before being merged, hence allowing for an appropriate estimation of the im-
143 pact of changes in vegetation cover. In our control simulation, the natural tile is han-
144 dled by the Noah-MP land surface model (Niu et al., 2011) and is considered as a mix-
145 ture of cropland and grassland. Lastly, BEP-BEM offers a set of dynamic models for ac-
146 tive and passive strategies for cooling outdoor and indoor environments, including green
147 roofs, solar PV panels, and air conditioning (Salamanca et al., 2011; Zonato et al., 2021).

148 We use the same model dynamics and physics set up as (Brousse et al., 2023) as
149 these have shown good agreement with observations for the whole summer 2018. For the
150 urban canopy parameters and the land cover classification we use the defaults MODIS
151 land cover from WRF and categorize the urban pixels to Local Climate Zones (LCZ; Stewart
152 and Oke (2012)) following the WUDAPT-TO-WRF strategy (Brousse et al., 2016; De-
153 muzere et al., 2022) to provide the WRF model with spatially explicit urban morpho-
154 logical parameters. The building characteristics are provided per LCZ and are the same
155 than in Brousse et al. (2023).

156 2.2 Model evaluation

157 To evaluate the model outputs in parts of the urban environment deprived of of-
158 ficial weather data sources, we follow the same evaluation strategy as in Brousse et al.
159 (2023) by taking advantage of the dense network of personal weather stations (PWS)
160 in the city of London (Brousse et al., 2022). Our model is thereby evaluated against of-
161 ficial automatic weather stations (AWS) from the MIDAS Met Office network (Sunter,
162 2021; UKMO, 2021) and quality-checked crowd-sourced PWS from the Netatmo com-
163 pany (<https://netatmo.org>). More details on the data acquisition and on the ration-
164 ale defending the use of privately owned PWS for UCM evaluation can be found in Brousse
165 et al. (2023). All our crowd-sourced weather data has been quality-checked and filtered
166 using a common statistical method provided by the *Crowd-QC* package from Grassmann
167 et al. (2018); also see Napoly et al. (2018).

We evaluate our model with several statistical indicators including the squared Pearson’s coefficient of correlation (r^2), Spearman’s coefficient of correlation, root mean squared error (RMSE), mean absolute error (MAE), and mean bias (MB). The evaluation is carried out over the whole domain to ensure that the model is performing well in all urban settings and to ensure that our choice for the GLA area is not biased towards higher model performances.

2.3 Temperature interventions

We define a set of 9 passive interventions that are theoretically applicable over the Greater London metropolitan area. Most of our interventions consist of roof interventions and each intervention is compared to a control run which consists of the same model set up as in Brousse et al. (2023), and which did not consider any heat intervention strategy. Therefore, the control run is simply representative of the impact of urban areas on the local climate while all subsequent scenarios estimate the impact of each intervention on that urban climate. We also simulate the effect of widespread use of active cooling in the form of air conditioning.

Our scenarios consist of: 1) the deployment of cool roofs by increasing the albedo of each urban class to 0.85 (CR_{100}); 2) the deployment of green roofs composed of sedum vegetation in all urban classes by activating the green roof model from Zonato et al. (2021) with its default irrigation hours from 23.00 UTC to 01.00 UTC (GR_{100}); 3) the deployment of solar PV panels on all urban roofs by activating the solar PV model from (Zonato et al., 2021) in its default version (PV_{100}); 4) the installation of air conditioning systems in all buildings and at all floors by activating the air conditioning model embedded in the BEM model in its default version (Salamanca et al. (2010); Salamanca and Martilli (2010); AC_{100}); and 5) the change of the urban vegetation from cropland and grass land to urban deciduous trees as implemented in NOAH-MP (Niu et al. (2011), VG_{100}). These scenarios aim to represent the maximum impact that one intervention can have on the city’s temperatures and SEB. More information about each model can be found in the appendix Appendix A.

We also test the impact of certain scenarios that are based on more achievable or tailored interventions. These scenarios are: 1) the potential for solar PV deployment over roofs of existing buildings in the Greater London taking into account roof slope and overshadowing derived from the London solar opportunity map (PV_{pot} ; Steadman et al. (2020)); 2) the potential for retrofit of green roofs on existing buildings in Greater London obtained through a detailed geographical information system analysis (C. Simpson et al. (2022); GR_{pot}); 3) the installation of cool roofs in compact and open mid-rise areas (Local Climate Zones 2 and 5) from the centre of London (CR_{mid}); 4) in open low rise areas (Local Climate Zone 6; CR_{low}) from the residential suburbs of London; and 5) in commercial and industrial areas of London (Local Climate Zone 8; CR_{cid}).

Potential conjoint use of green and cool roofs with solar panels have been proposed to increase solar panel efficiency while reducing urban surface temperatures in particular (Fleck et al., 2022; Schindler et al., 2018), but it was not possible to analyse this in our model. Sedum was modelled for the green roofs due to its increasing presence on new buildings in London (C. Simpson et al., 2022), but other plants such as turf may have a different impact.

To integrate the spatially-explicit variation in solar PV panels and green roofs coverage from the scenarios PV_{pot} and GR_{pot} , we adapted the code from Zonato et al. (2021) and added a field to the WRF geographical input files. These two scenarios also explain why our study focuses only on GLA area as these are bounded to its administrative boundaries and are not applicable to the whole south-east of England which composes our WRF domain. Lastly, we expect the impacts to be more prominent in London due to its higher urban heat island intensity (Bassett et al., 2021; Brousse et al., 2022).

3 Results

During the two hottest summer days of 2018, our model configuration presents a warm bias estimated at $0.72\text{ }^{\circ}\text{C}$ on average in urban PWS and of $1.76\text{ }^{\circ}\text{C}$ amongst official AWS (Fig. B1). Such a difference may be explained by the fact that official AWS are often located in open fields like parks or airports, which further supports the use of PWS for model evaluation (Hammerberg et al., 2018; Brousse et al., 2023). Metrics estimating the model error are better on average across urban PWS compared to AWS, with average values of $2.39\text{ }^{\circ}\text{C}$ compared to $2.84\text{ }^{\circ}\text{C}$ for RMSE, and of $1.94\text{ }^{\circ}\text{C}$ compared to $2.33\text{ }^{\circ}\text{C}$ for MAE. The model seems however to be better correlated temporally at AWS locations with coefficients of 0.91 against 0.88 for Spearman’s r and of 0.84 against 0.81 for Pearson’s r^2 . We consider our model performance to be acceptable, especially considering the high level of confidence in the temporal evolution of the air temperature at 2 m (T_2) which is directly correlated to the SEB. We thereby assume that our model is trustworthy for each SEB component as we cannot evaluate them without access to flux tower measurements.

In general, we find that cool roofs are the more effective interventions to reduce the heat in the GLA boundaries with a 2-days average cooling of $\sim 1.2\text{ }^{\circ}\text{C}$ going up to $\sim 2.0\text{ }^{\circ}\text{C}$ in certain locations (Fig. 1 and 2). All cool roofs interventions contribute to a reduction of T_2 on average (Fig. 2) with major impacts found in the south and the east of GLA if deployed at 100 % (CR_{100} ; Fig. 1). Horizontal advection of cooling depends on the location where cool roofs are implemented. For instance, if cool roofs are implemented in more central and mid-rise areas of London, more open and low-rise residential areas appear to benefit from this cooling by $\sim 0.15\text{ }^{\circ}\text{C}$ on average. Interestingly, implementing cool roofs on residential low-rise buildings only (CR_{low}) does not cool the central parts of the cities despite providing a substantial outdoor T_2 reduction by up to $\sim 2\text{ }^{\circ}\text{C}$ in the south-east. Establishing cool roofs only on industrial and commercial areas (CR_{cid}) only results in a marginal heat reduction on average over London but does lead to a heat reduction of up to $\sim 1\text{ }^{\circ}\text{C}$ in large industrial areas of east-London. Compared to cool roofs, we find more minor decreases in T_2 of $\sim 0.3\text{ }^{\circ}\text{C}$ on average related to both the change in natural cover from a mixture of croplands and grasslands to deciduous trees (VG_{100}) and to the installation of solar panels over the whole city (PV_{100}). Stronger cooling effects are found in the east of London for these two scenarios. Lastly, despite a high spatial heterogeneity of the cooling provided by the remaining interventions (green roofs (GR_{100} and GR_{pot}); realistic retrofit potential of solar panels (PV_{pot}), none is found to be beneficial for T_2 reduction on average for the whole metropolitan area (Fig. 2). Using air conditioning for maintaining the indoor temperature of all London’s building stock to $21\text{ }^{\circ}\text{C}$ (AC_{100}) induces a heating of central London by up to $\sim 1\text{ }^{\circ}\text{C}$ and of $\sim 0.15\text{ }^{\circ}\text{C}$ on average for the whole city.

By relating the change in T_2 of all urban pixels in the GLA area to their relative SEB components we find that the large decrease in T_2 induced by the full scale deployment of cool roofs (CR_{100}) is linked to an increase in albedo by ~ 0.3 and a reduction of sensible heat fluxes emitted by the city of $\sim 30\text{ }W\cdot m^{-2}$, on average. Notably, however, we observe an increase in incoming solar radiation when deploying cool roofs over the whole city (CR_{100}) and in open residential areas (CR_{low}) as well as when implementing solar panels over the whole city (PV_{100}); something that was also recently observed by Valencia et al. (2023). This could be related to a lower convection due to the cooler surfaces, although major changes in the average cloud cover in the whole atmospheric column are not perceived. Higher emissions of sensible heat fluxes by $\sim 5\text{ }W\cdot m^{-2}$ are found in both the air conditioning (AC_{100}) and the deciduous tree (VG_{100}) scenarios. While the former is expected as AC systems remove the accumulated indoor heat outside mostly in the form of sensible heat, the latter is counter intuitive as more vegetated areas are expected to mostly transform energy in the form of latent heat fluxes via evapotranspiration – something that is modelled with an average increase of latent heat fluxes

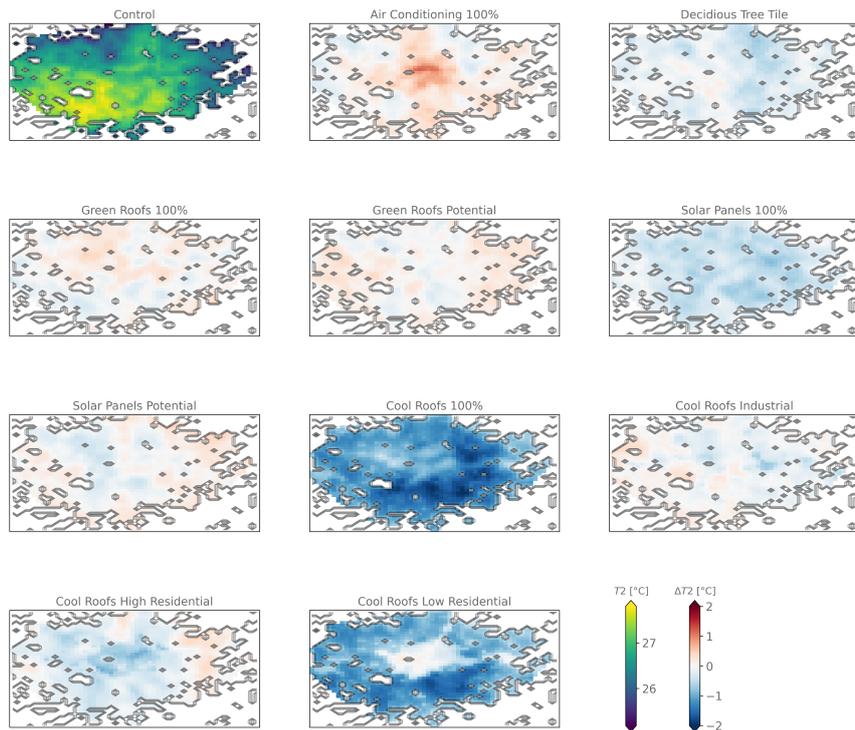


Figure 1. Average 2 m air temperature of the control run (T_2 ; top left panel) and the average difference of each intervention (ΔT_2). Grey contours are WRF urban pixels (classified as an urban LCZ).

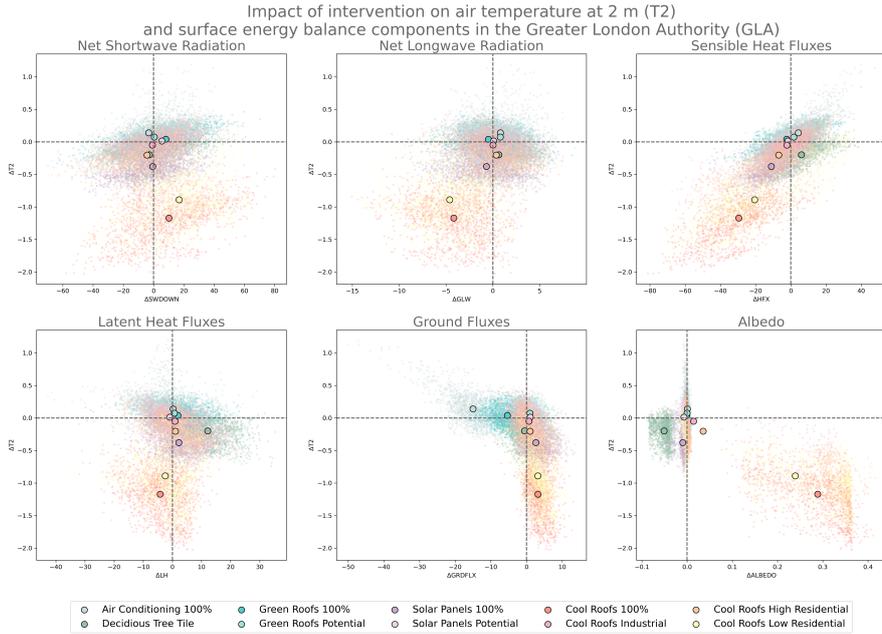


Figure 2. Time-averaged impact of each intervention on each surface energy balance component and the respective change in air temperature at 2 m for all urban pixels (small dots). Average impact is represented by the large dots.

272 by $\sim 12 \text{ W}\cdot\text{m}^{-2}$. We however observe a decrease in surface albedo by 0.05 on average
 273 in parallel which could explain this phenomenon.

274 Despite an increase of T_2 induced by green roofs of less than $0.1 \text{ }^\circ\text{C}$ on average,
 275 hourly spatial averages of changes in T_2 show that if green roofs were implemented across
 276 the whole city (GR_{100}), temperatures could be decreased by ~ 0.5 to $\sim 0.8 \text{ }^\circ\text{C}$ during hot
 277 hours of the day (11.00 UTC to 14.00 UTC; Fig. 3 and Fig. C1 to C2). Nonetheless, temper-
 278 atures would also be increased at times of daily minimum T_2 by $\sim 0.5 \text{ }^\circ\text{C}$ (03.00 UTC
 279 to 06.00 UTC, Fig. 3). This is consistent with findings by Zonato et al. (2021) and ex-
 280 plained by an increased release of latent fluxes during the day but a higher storage of
 281 heat fluxes at night due to green roofs' thermal mass. The current potential for green
 282 roofs in London (GR_{pot}) would not reduce the temperature by more than $0.1 \text{ }^\circ\text{C}$ on av-
 283 erage across the period. During the hotter hours of the day, the impact of cool roofs (CR_{100})
 284 can be 2.5 times and the one from cool roofs over open-low rises (CR_{low}) 2.0 times greater
 285 than the one of green roofs. Both these scenarios (CR_{100} and CR_{low}) also contribute to
 286 a constant reduction of temperature of at least $0.5 \text{ }^\circ\text{C}$ throughout the day. This is cor-
 287 related to a high reduction of emitted sensible heat fluxes and of the positive intakes of
 288 ground fluxes throughout daytime hours (Fig. C1 to C2). Notably, the temperature re-
 289 duction across the whole city induced by cool roofs over central mid-rises from London
 290 (CR_{mid}) reaches a peak at later hours, around 15.00 UTC, a few hours after the peak in
 291 reduction of sensible heat fluxes reached at noon. Higher temperature reduction by
 292 $\sim 0.8 \text{ }^\circ\text{C}$ is observed in the evening if vegetation is changed to deciduous trees (VG_{100}).
 293 During the day, no tangible impact is found in this case, as half of the urban pixels ob-
 294 serve an increase in temperature while the other half observes a decrease. This may be
 295 because higher levels of sensible heat fluxes are emitted; more latent heat fluxes are also
 296 emitted throughout the day. Both full scale and realistic solar panel implementation (PV_{100}

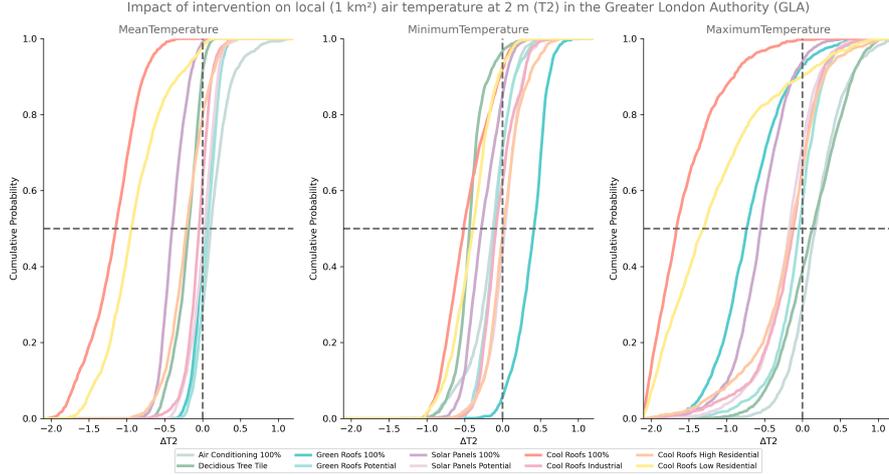


Figure 3. Cumulative distributions of change in air temperature at 2 m for different intervention scenarios. Distributions are of change relative to the reference scenario in average daily mean, daily minimum, and daily maximum 2 m air temperature across grid cells. Distributions are from 200 bin histograms, normalised to unity and based on modelled air temperature on 26th and the 27th of July 2018 in London. The horizontal dashed line indicates the median change. The vertical dashed line indicates zero change.

297 and PV_{pot}) behave similarly to cool roofs by decreasing the emitted sensible heat fluxes
 298 and reducing the intake of fluxes to the ground during daytime hours. The temperature
 299 reduction of up to ~ 1.0 °C, for the full-scale deployment, however, only happens dur-
 300 ing the late afternoon and the evening. At time of daily maximum temperatures, PV_{100}
 301 reduces the temperature in similar ways to GR_{100} (3 and Fig. C1 to C2).

302 4 Discussion

303 Using the WRF BEP-BEM mesoscale urban climate over the Greater London metropoli-
 304 tan area during the two hottest days of 2018, we performed one of the first comprehen-
 305 sive evaluations of the impact of building and street level interventions tailored at re-
 306 ducing outdoor and indoor heat on the outdoor temperature and the surface energy bal-
 307 ance in London.

308 We find that interventions such as cool roofs, green roofs, rooftop solar PV, and
 309 changing natural spaces to deciduous woodland can contribute to a cooling of outdoor
 310 temperatures at hyper-local or city scales (also see Fig. C3 to C8, while widespread use
 311 of air conditioning could lead to substantial increases in outdoor temperature in the cen-
 312 tre of London. On average, cool roofs would be the most effective way of reducing out-
 313 door temperature at the city scale, in line with other modelling studies (Sharma et al.,
 314 2016; Yang & Bou-Zeid, 2019; Tan et al., 2023). At the maximum, this temperature re-
 315 duction is between 3.2 °C and 2.8 °C at times when daily spatial average temperatures
 316 reached 33 °C and 37 °C, respectively (Fig. C2), similar to the cooling found in Phoenix
 317 and New York (Sinsel et al., 2021). In comparison, if green roofs or rooftop solar PV pan-
 318 els were deployed over all buildings in the Greater London area, the temperature reduc-
 319 tion would have never reached more than ~ 1.0 °C for both days; considering that overnight,
 320 green roofs would increase the average temperature due to their increased heat-storage
 321 capacity that increases heat release at night (Tan et al., 2023). Of note, our green roofs

322 are modelled with sedum vegetation as they are the most common in London; other de-
323 signs may lead to other effects as Zonato et al. (2021) found an increased reduction us-
324 ing grass-based green roofs. Conversely, we find that changing the modelled vegetation
325 type of urban natural spaces from grasslands/croplands to deciduous trees would reduce
326 outdoor temperatures only during the night and found mixed effects during the day, pos-
327 sibly due to an increase in the surface albedo which leads to an increased sensible heat
328 emission during the day. Latent heat would also increase as expected from more evap-
329 otranspiration and would be equivalent to the increase in latent heat from green roofs
330 deployed at full scale. Although cooling via evapotranspiration reduces air temperature,
331 it increases water vapor in the atmosphere; this could be beneficial for avoiding water
332 stress of vegetation but affect the heat stress from city's inhabitants and should be eval-
333 uated using a set of multiple heat stress metrics (C. H. Simpson et al., 2023).

334 City scale impacts of any intervention are reduced when applied to a smaller area
335 or fewer buildings. For cool roofs, we find that although major impacts are found when
336 applied over low-rises residential areas of London, already applying them over central
337 mid-rises of the city could reduce the temperature in surrounding residential areas. Be-
338 cause of the relatively small area covered by industrial areas in London, applying cool
339 roofs over them would only marginally contribute to a heat reduction of the city, con-
340 tradictory to results in Birmingham where commercial and industrial areas cover an im-
341 portant part of the city centre (H. Macintyre & Heaviside, 2019). They could also be con-
342 sidered less effective as industrial buildings already observe higher albedo than more res-
343 idential areas (Brousse et al., 2023). Cool roofs may act to reduce indoor temperatures
344 in summer, but also potentially in winter, thus increasing winter heating demand. There-
345 fore, it is important that they are installed in conjunction with insulation in homes (Taylor
346 et al., 2018). However, in winter, cool roofs have less of a cooling effect outdoor temper-
347 atures than they do in summer (H. L. Macintyre et al., 2021). While we find that full
348 application of rooftop solar PV leads to some cooling on average, realistic application,
349 constrained by existing building characteristics, is not enough to have an appreciable ef-
350 fect on the temperature. Similarly, full application of sedum green roofs can reduce day-
351 time temperatures, but the potential for retrofit of existing buildings with sedum green
352 roofs is not enough to produce a substantial cooling effect at the city scale.

353 Although air conditioning would increase the outdoor temperature, it could still
354 protect vulnerable populations from overheating indoors. Through our modelling exper-
355 iment, we find that if solar panels were implemented in a realistic manner on the rooftops
356 of existing buildings, the energy production, peaking at ~ 22.5 MW at time of higher solar
357 incidence (11.00 UTC) and with an average production of ~ 7 MWh, would cover the
358 energetic consumption induced by AC systems which never goes above ~ 9 MW during
359 the hottest hours (15.00 UTC to 17.00 UTC), averaging to ~ 4.5 MWh. Under the as-
360 sumption that deployment of solar panels is not constrained by existing buildings, elec-
361 tric power production is 75 % greater than the realistic retrofit scenario, and outdoor
362 average temperature is reduced by ~ 0.3 °C while temperature reduction is negligible in
363 the realistic case. This temperature reduction should be carefully interpreted as several
364 studies have found that solar panels could heat up the surrounding environment (Sailor
365 et al., 2021). Nonetheless, these local impacts may be dependent on many factors, in-
366 cluding the building height and the building plan area fraction (Zonato et al., 2021), or
367 the increase in rooftop albedo. In our case, we argue that the capture of solar radiation
368 by the solar PVs is sufficiently important to lower the expected intake of radiation by
369 the building and that most of it is transformed in the form of electric energy rather than
370 in sensible heat over the roof. In the evening and at night, the low thermal capacity of
371 solar PVs explains the higher cooling capacity of the urban environment. In any case,
372 our results call for more investigation on this specific topic as other studies, like (Tan
373 et al., 2023), found a small night-time increase in temperature over Chicago when im-
374 plementing solar PV over buildings' roofs. Hence mechanisms explaining solar PV ef-
375 ficiency in our model need to be further explored.

376 Several limitations apply to this study. First, we used only one UCM, and another
377 UCM may produce different results depending on its dynamics and physics (Krayenhoff
378 et al., 2021). Second, we focused our study only over two hot summer days due to com-
379 putational limitations. In general, performing such comprehensive analysis with multi-
380 ple scenarios comes at high computational costs, which often prevents the generalization
381 of the results to expected average seasonal impacts of these interventions or to other cities
382 at similar latitudes. Third, despite the fact that the 26th and the 27th of July were ex-
383 tremely hot days during the summer 2018 in the United Kingdom, full clear-sky condi-
384 tion is absent from our model simulations, which could, for instance, partially explain
385 why solar PV do not lead to a heating of the city during the day. Some of the results
386 observed in this study concerning the surface energy balance may therefore simply be
387 related to the overpass of clouds; although we did not find a particular difference in cloud
388 coverage between our control simulation and our sensitivity tests. Fourth, as discussed
389 above, our study only focuses on outdoor temperatures and does not consider the ben-
390 efits or the detriments caused by these interventions on other aspects (e.g., biodiversity,
391 indoor temperature, increase in relative humidity, wind circulation, etc.). Lastly, all our
392 estimated impacts should be interpreted cautiously as the average temperature differ-
393 ences resulting from the interventions are smaller than the model MAE with respect to
394 observations (Krayenhoff et al., 2021). Despite these limitations, we expect our results
395 to be at the least indicative of which interventions are capable of producing the great-
396 est cooling effect in London.

397 Notwithstanding, the outcomes of this study also have to be put in perspective against
398 the costs related to the deployment of each of these interventions which should be weighted
399 according to a cost-benefit analysis.

400 5 Conclusions

401 Adaptation to urban heat is of increasing priority. Our modelling study suggests
402 that widespread adoption of cool roofs could be efficient at reducing urban temperatures,
403 and therefore adverse health impacts, during hot spells. Rooftop solar PV may also pro-
404 vide a small amount of cooling as an additional benefit to power production; green roofs
405 appear less effective in reducing temperatures but have other environmental benefits. Fur-
406 thermore, relying on AC will lead to increased outdoor temperatures, despite improved
407 thermal comfort indoors. It is reasonable to think that there will be a mixture of adap-
408 tation measures implemented across the city of London, and their applicability and im-
409 pact should be further investigated. Our research is a first step towards this and will be
410 of interest to city planners and decision-makers within the built environment sector.

411 6 Open Research

412 The simulations done in this research were performed using the WRF model v4.4
413 (<https://github.com/wrf-model/WRF.git>). The related outputs presented in this re-
414 search and codes used to plot them are available following this DOI: 10.5281/zenodo.8333363

415 Acknowledgments

416 CH is supported by a NERC fellowship (NE/R01440X/1) and acknowledges funding for
417 the HEROIC project (216035/Z/19/Z) from the Wellcome Trust, which funds OB and
418 CS.

419 OB designed the study and led the conception of the manuscript with the support
420 of CH and CS. OB was responsible for the WRF modelling, the model evaluation and
421 the analysis. CS provided support in the integration of green roofs and solar PV inputs
422 into WRF. AZ implemented a spatially-explicit input of solar PV and green roofs cov-
423 erages in WRF. AZ and AM offered guidance in the set-up of the WRF model v4.4 and

424 in the analysis of the outputs. JT and MD provided support in the design of the study
425 and in the acquisition of knowledge about building interventions. All authors contributed
426 to the writing of the manuscript.

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598 **Appendix A Model parameterization for interventions**

599 In this study, the model parameterization follows exactly the same as in Brousse
600 et al. (2023). This means that the model is run at a resolution of 1 x 1 km with 210 by
601 180 horizontal grid points and 71 vertical layers nested in two other domains run at 12
602 and 3 km, respectively, and forced by ERA5 6-hourly data at 25 km horizontal resolu-
603 tion. We supplement the version 4.3 to the more recent 4.4.1 to benefit for certain bug
604 fixes; these did not relate to the urban scheme. Of importance, sea surface temperatures
605 are updated every 6 hours out of ERA5, no lake models are activated for inland water
606 bodies, and initial land surface conditions are provided by the default MODIS 5-arc-second
607 land use data set. The latter are further interpolated by the Noah-MP land surface model
608 (Niu et al., 2011) in its default parameterization over 4 soil layers. Building effects on
609 the local climate are calculated by activating the BEP-BEM models (Martilli et al., 2002;
610 Salamanca et al., 2010; Salamanca & Martilli, 2010) which calculates energy fluxes within
611 the urban tile of each grid points – given by the urban fraction. Changes in urban veg-
612 etation like in VG₁00 are therefore only treated by Noah-MP within the natural tile of
613 grid points where the urban fraction is not null. Fluxes are then averaged at the bulk
614 level to estimate prognostic variables at the each grid point (e.g., air temperature). We
615 chose to use the common Bougeault-Lacarrère planetary boundary layer scheme for these
616 simulations (Bougeault & Lacarrere, 1989). More information on the physical param-
617 eterization and on the buildings’ thermal and radiative properties can be found in Brousse
618 et al. (2023) under the section 2.a.

619 The impact of the air conditioning in BEP-BEM is estimated by means of a sim-
620 ple Building Energy Module (BEM). This module computes an energy budget of the in-
621 door air by considering the heat generated by people and equipment, the diffusion of heat
622 through walls and roof, the air infiltration/ventilation, and the radiation entering through
623 the windows. When the indoor air temperature reaches maximum value fixed by the user,
624 the internal temperature is kept constant, and all the extra heat (H_{sneed}) is ejected to
625 the atmosphere. In addition to the atmosphere is added also the heat generated by the
626 A.C. equipments to do the work (H_{sneed}/COP , where COP is the Coefficient of per-
627 formance of the A.C. system). In the same way the energy consumption due to the A.C.
628 is estimated as H_{sneed}/COP . In our AC₁00 simulations, all buildings are equipped with
629 AC systems running with a COP of 3.5 and a target temperature of 294.15 K. More de-
630 tails can be found in Salamanca et al. (2010); Salamanca and Martilli (2010).

631 The roof mitigation strategies parameterizations are based on the work of Zonato
632 et al. (2021). The land-surface scheme for green roofs has been developed based on De Munck
633 et al. (2013). It calculates energy and water budgets, taking into account incoming net
634 radiation, water input from precipitation and irrigation (the latter considered as irriga-
635 tion at the green roof surface), evapotranspiration from vegetation, heat exchange with
636 the atmosphere, and diffusion of energy and moisture throughout the soil. A green roof
637 consists of 10 layers with a total depth of 0.3 m, 5 of them represents the vegetation and
638 the soil substrates, while the rest the underlying roof, including the waterproof mem-
639 brane. The kind of vegetation present in the upper level is parameterized depending on
640 leaf area index and maximum stomatal resistance.

641 The parameterization taking into account the effects of RPVPs assumes the pho-
642 tovoltaic arrays to be parallel and detached from roofs and composed of a single layer,
643 and it is based on the work of Jones and Underwood (2001). A photovoltaic panel is as-
644 sumed to be detached from the roof, and to composed of three layers, as in: a monocrys-
645 talline silicon PV cell, a polyester trilaminate and a glass face, with a total depth of 6.55
646 mm. The prognostic equation of its temperature, that is necessary for calculating the
647 incoming/outgoing heat fluxes, considers: 1) The net incoming short- and long-wave ra-
648 diation at both surfaces of the photovoltaic panel, assuming a view factor for the bot-
649 tom surface depending on the area covered by the photovoltaic panel and on its distance
650 to the underlying roof; 2) The heat fluxes dependent on wind speed and temperature dif-

651 ferences between the panel and the air (Scherba et al., 2011); 3) The energy produced
652 by the PV cell, dependent on its efficiency and on the PV temperature itself.

653 The standard WRF version 4.4.1 has been appropriately modified in order to con-
654 sider a grid-specific ratio of green roof or photovoltaic panels independent of the LCZ,
655 thus independent of look-up tables.

656 **Appendix B Observational Data and Model Evaluation**

657 The model evaluation was performed following the strategy described in Brousse
 658 et al. (2023) and uses the same data set. Briefly put, personal weather station air tem-
 659 perature measurements from the *Netatmo* company are gathered using an open API. Each
 660 personal weather station's data undergoes a statistical quality-check to ensure that the
 661 quality of the measurement is sufficient to perform urban climate studies and model eval-
 662 uation (Napoly et al., 2018). These crowd-sourced measurements complement the offi-
 663 cial weather stations measurements coming from the UK MetOffice MIDAS network (UKMO,
 664 2021). Hammerberg et al. (2018) and Brousse et al. (2023) indeed demonstrated that
 665 crowd-sourced weather data are beneficial for evaluating urban climate simulations and
 666 we therefore decided to take advantage of them in this study too. More information on
 667 the data gathering and treatment can be found in Brousse et al. (2023) and related codes
 668 can be found following https://github.com/oscarbrousse/JAMC_BiasCorrection_PWS/.

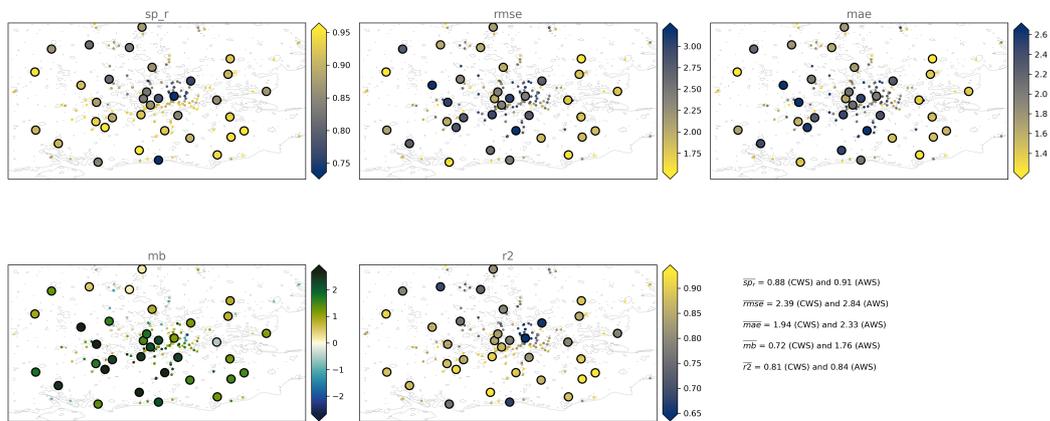


Figure B1. Model evaluation against Netatmo personal citizen weather stations (CWS, small dots) and MIDAS official automatic weather stations (AWS, big dots). Yellow is better. For MB white is the better. Average scores amongst all stations are given in the bottom right.

669 **Appendix C Additional Figures**

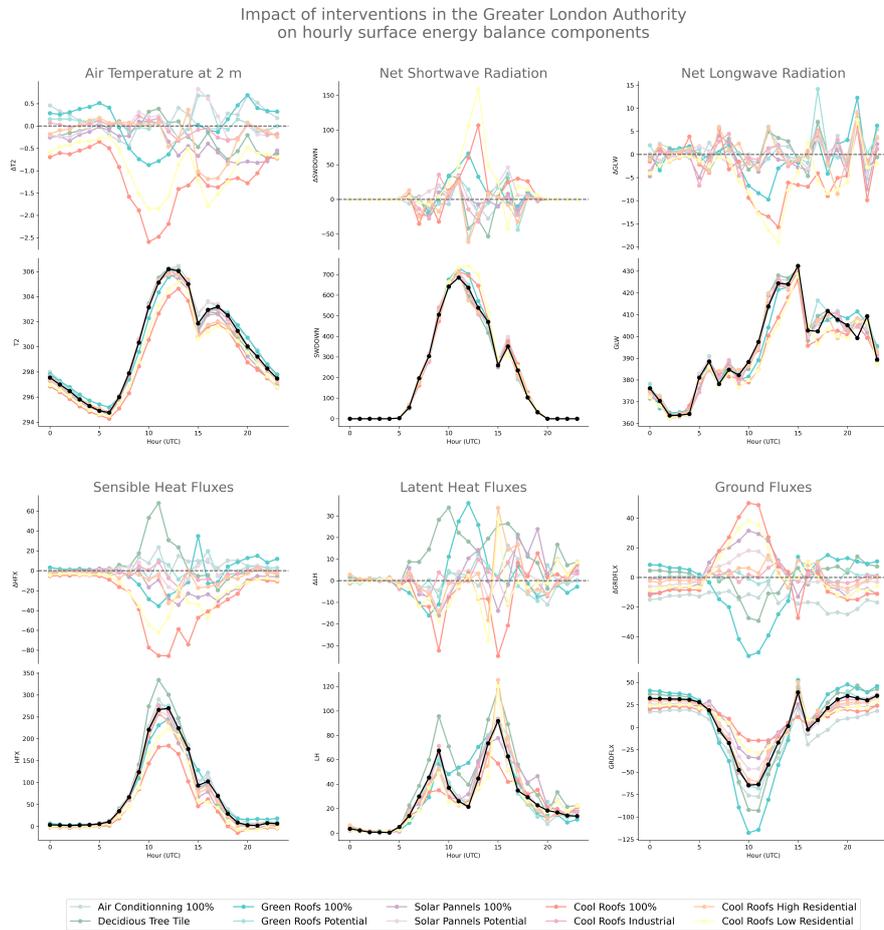


Figure C1. Diurnal cycles of spatially average air temperature at 2 m and of each surface energy balance component (lower panels) and their respective differences (control - intervention; upper panels) over GLA . The black solid line is the control run, the interventions' colors are given in the legend. The dashed grey line represents a null change between the control and the intervention run.

Impact of intervention on hourly surface energy balance components

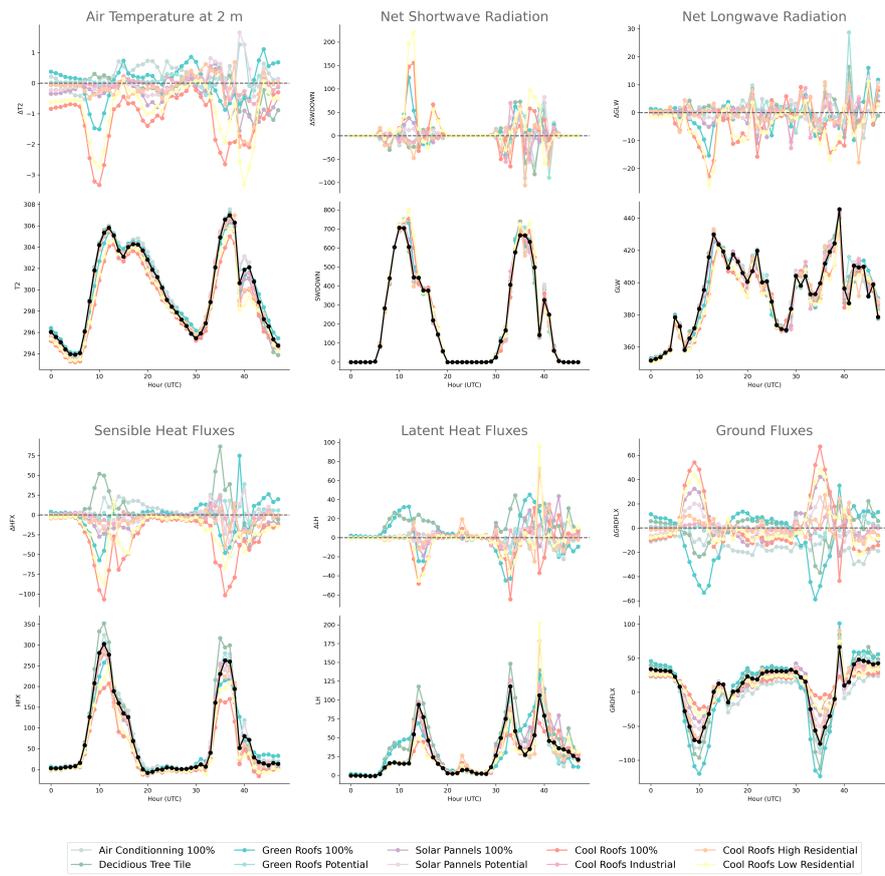


Figure C2. Same as Fig. C1 but without hourly averaging (for the two full days)

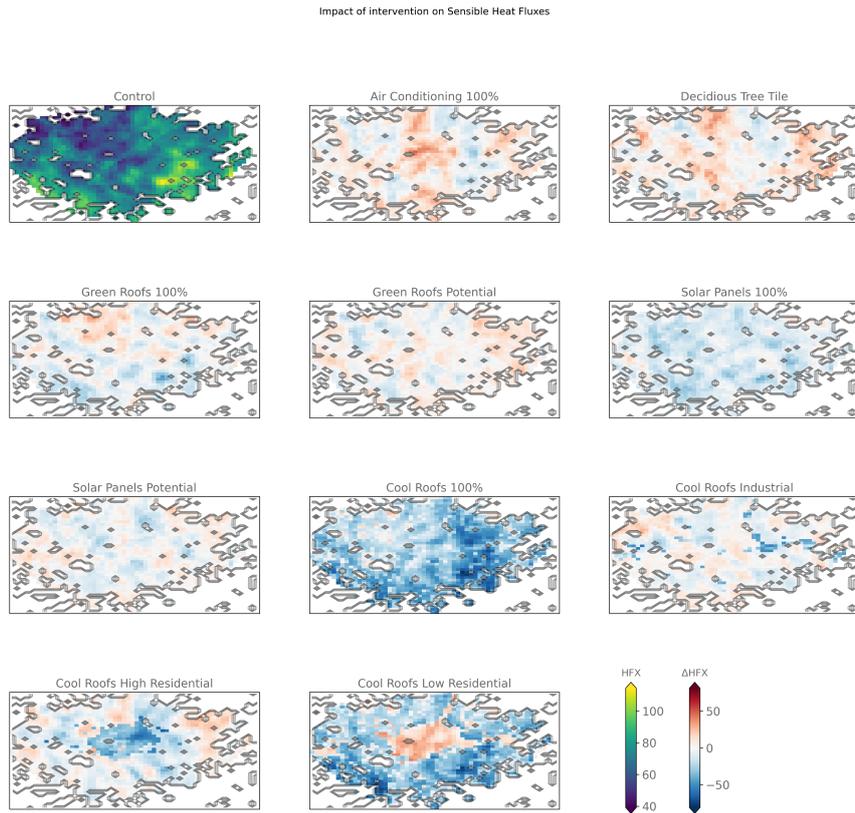


Figure C3. Same as Fig. 1 but for sensible heat fluxes

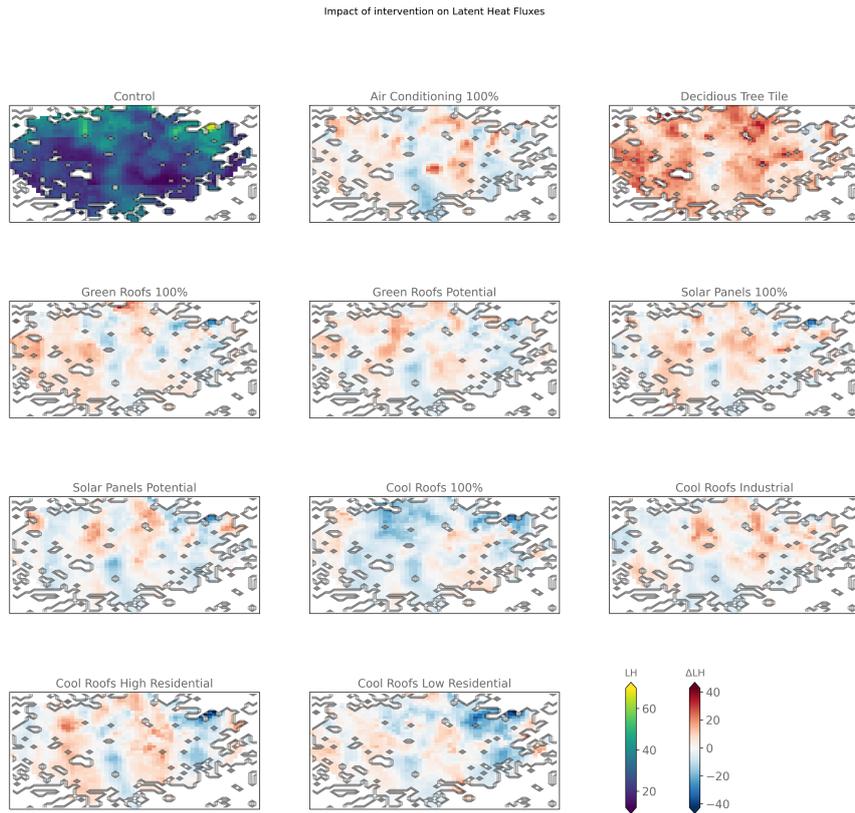


Figure C4. Same as Fig. 1 but for latent heat fluxes

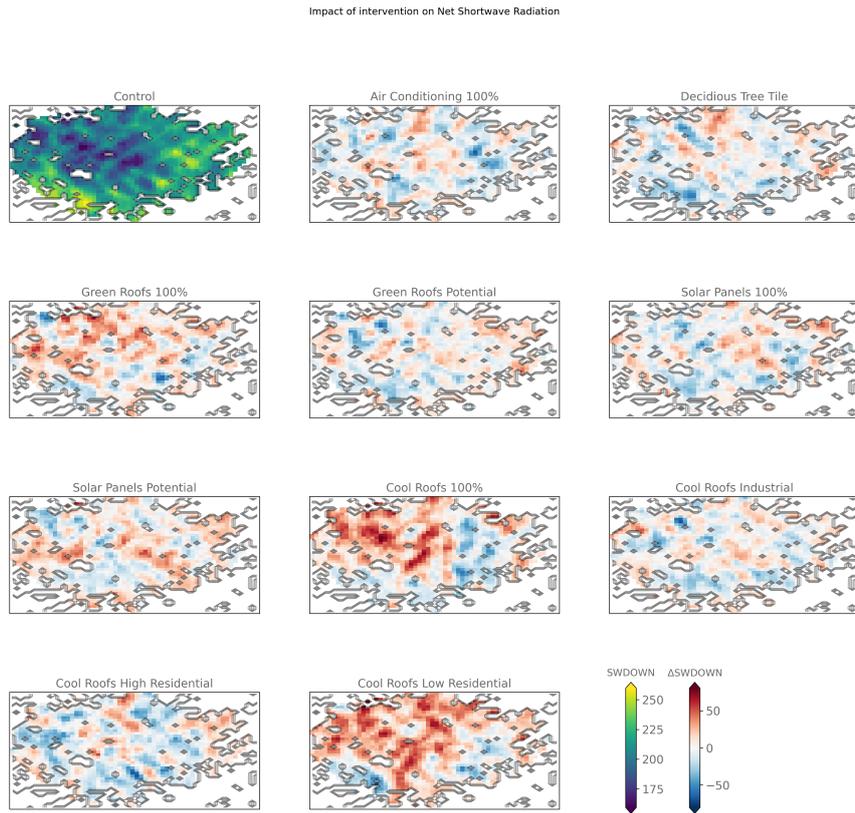


Figure C5. Same as Fig. 1 but for net incoming short-wave solar radiation

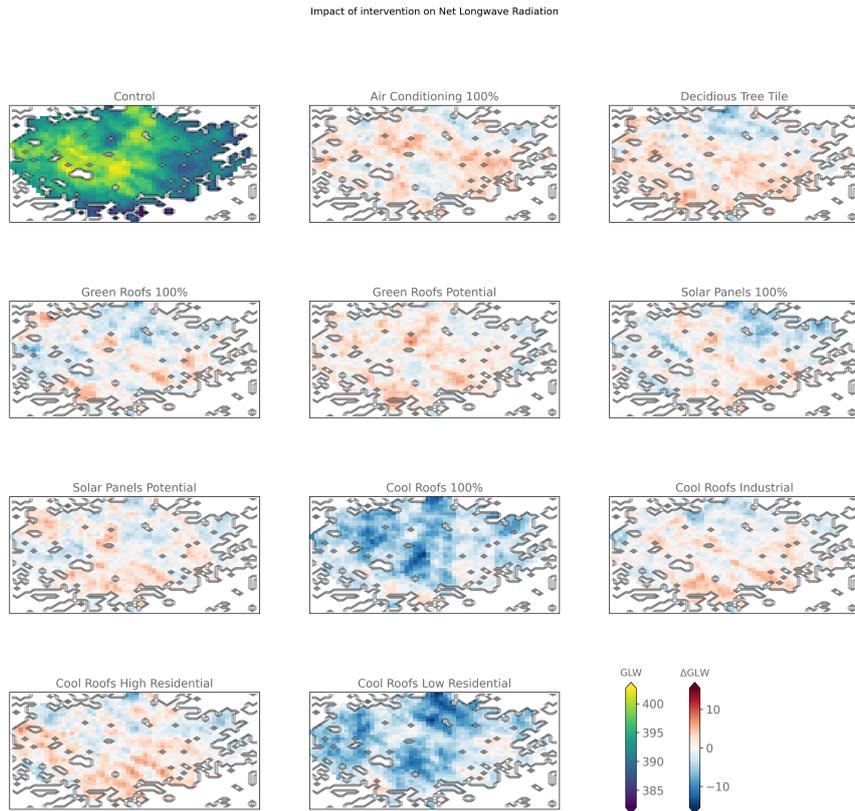


Figure C6. Same as Fig. 1 but for net incoming long-wave radiation

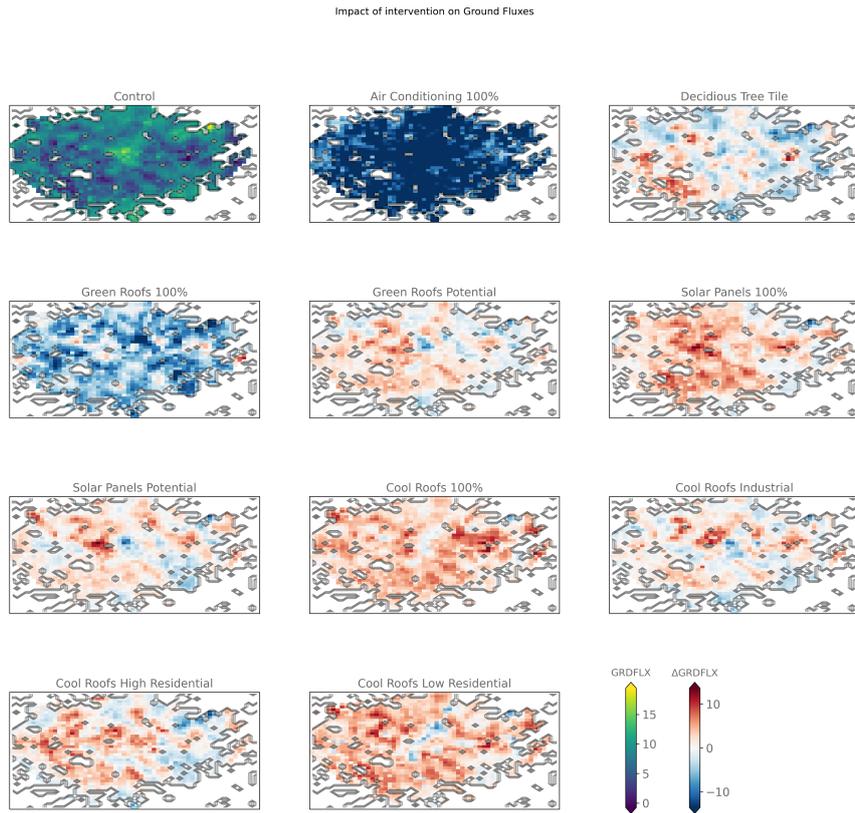


Figure C7. Same as Fig. 1 but for ground fluxes

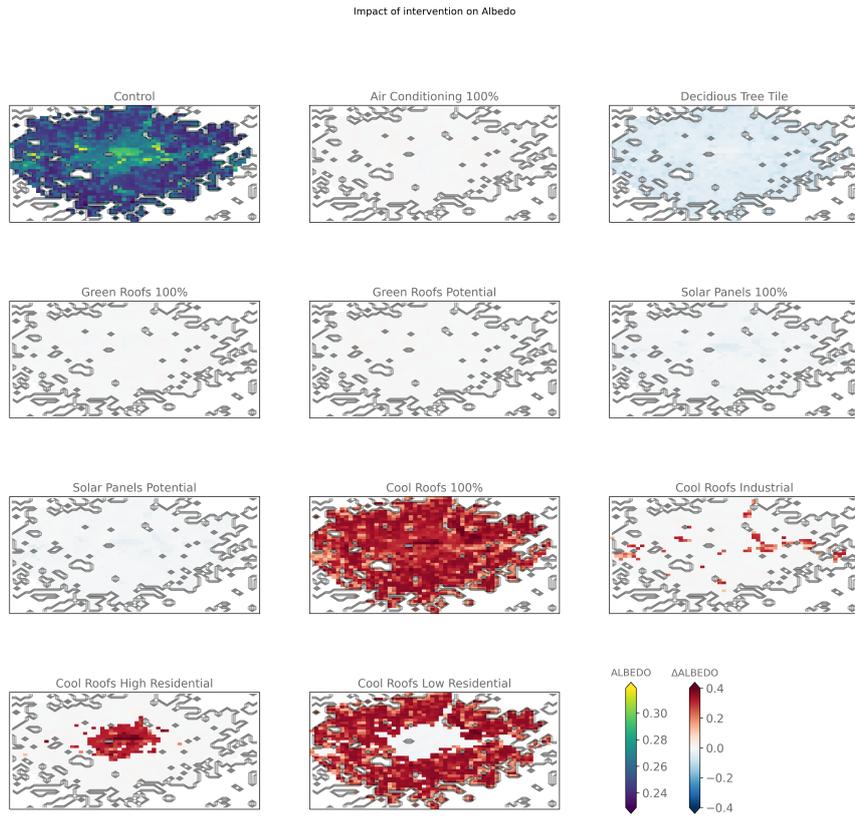


Figure C8. Same as Fig. 1 but for albedo