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Xiangpu Gong^{1,2}, Nicole Itzkowitz³, Calvin Jephcote¹,
Kathryn Adams¹, Glory O Atilola³, John Gulliver^{1,2},
Marta Blangiardo³ and Anna Hansell^{1,2*}

¹Centre for Environmental Health and Sustainability, University of Leicester,
Leicester, UK

²The National Institute of Health and Care Research (NIHR) Health Protection
Research Unit in Environmental Exposure and Health at the University of Leicester,
Leicester, UK

³Medical Research Council Centre for Environment and Health, Department of
Epidemiology and Biostatistics, School of Public Health, Imperial College London,
London, UK

*Corresponding author

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Abstract

Impact of short-term aircraft noise on cardiovascular disease risk in the area surrounding London Heathrow airport: the RISTANCO epidemiological study

Xiangpu Gong^{1,2}, Nicole Itzkowitz³, Calvin Jephcote¹,
Kathryn Adams¹, Glory O Atilola³, John Gulliver^{1,2},
Marta Blangiardo³ and Anna Hansell^{1,2*}

¹Centre for Environmental Health and Sustainability, University of Leicester, Leicester, UK

²The National Institute of Health and Care Research (NIHR) Health Protection Research Unit in Environmental Exposure and Health at the University of Leicester, Leicester, UK

³Medical Research Council Centre for Environment and Health, Department of Epidemiology and Biostatistics, School of Public Health, Imperial College London, London, UK

*Corresponding author ah618@leicester.ac.uk

Background: Long-term exposure to aircraft noise has been associated with small increases in cardiovascular disease risk, but there are almost no short-term exposure studies.

Objectives: Research questions were:

Is there an association between short-term changes in exposure to aircraft noise and cardiovascular morbidity and mortality?

What are the key effect modifiers?

Is there variability in risk estimates between areas with consistent versus changing patterns of noise exposure?

Do risk estimates differ when using different noise metrics?

Design: Descriptive analyses of noise levels and variability at different times of day, analyses of inequalities in noise exposure and case-crossover analyses of cardiovascular events in relation to aircraft noise exposure.

Setting: Area surrounding London Heathrow airport.

Time period: 2014–18.

Participants: Whole population in study area.

Main outcome measures: Cardiovascular disease hospitalisations and mortality.

Data sources:

Aircraft noise levels modelled using a standard noise model for: (1) daily equivalent continuous sound levels at different times of day; (2) daily number of events above defined noise thresholds (2018 only). National Health Service digital hospital admission records and Office for National Statistics mortality records for 2014–18 for cardiovascular outcomes, plus individual-level confounders available from healthcare records.

Confounder data including road traffic noise (Leicester modelled), rail noise and air pollution (Department for Environment, Food and Rural Affairs), area level deprivation and ethnicity (UK Census).

Results: The morning shoulder period (06.00–07.00 hours) was the noisiest of all eight bands (mean: 50.92 dB). The morning shoulder period also had the third highest number of noisy events (flights) > 60 dB per day, with three events across postcodes on average. However, the highest number of noisy events occurred in daytime (highest between 07.00 and 15.00 hours, second highest 15.00 and 19.00 hours).

To identify areas with high variability in aircraft noise exposure (due to changes in flight paths because of wind direction and airport operations), we used coefficients of variation (CoV). The period 24.00–04.30 hours had the highest mean CoV (67.33–74.16), followed by 04.30–06.00 hours and 23.00–24.00 hours.

Postcodes in the least deprived quintiles of Carstairs index or avoidable death rate had the lowest noise levels.

In case-crossover analyses, we observed increased risk for cardiovascular disease hospital admissions for evening noise 19.00–23.00 hours (odds ratio 1.005, 95% confidence interval 1.000 to 1.010 per 5 dB), but not for other periods or mortality. Further analyses suggested that increased risks were occurring in postcodes with low CoV for noise. We found effect modification by age, sex, ethnicity, deprivation and season.

Limitations: The industry standard noise model, the Aviation Environmental Design Tool, used does not take account of wind direction, which may have led to some exposure misclassification.

Conclusions: We developed a comprehensive dataset of daily aircraft noise variability.

We found small associations between cardiovascular hospitalisations (but not deaths) and evening aircraft noise levels, particularly in areas with low variability of noise.

Future work: More studies are needed to understand the effect of noise variation and respite/relief on cardiovascular disease.

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List of abbreviations

AEDT	Aviation Environmental Design Tool	Leve	equivalent continuous sound pressure level during 19.00–23.00 hours
CAA	Civil Aviation Authority	Lnight	equivalent continuous sound pressure level during 23.00–07.00 hours
CHD	coronary heart disease	LSOA	lower layer super output areas
CI	confidence interval	N60	number of flights > 60 dB in night-time
COA	census output areas	N65	number of flights > 65 dB in daytime
CoV	coefficients of variation	NHS	National Health Service
CVD	cardiovascular disease	NIHR	National Institute for Health and Care Research
dB	decibel	OR	odds ratio
EMAT	Early Morning Arrival Trial	PM _{2.5}	particulate matter less than 2.5 µm in diameter
ICD-10	International Classification of Diseases Version 10	SD	standard deviation
IHD	ischaemic heart disease		
LAD	local authority district		
LAeq24	equivalent continuous sound pressure level during 24 h		
Lday	equivalent continuous sound pressure level during 07.00–19.00 hours		

Plain language summary

Previous studies have found links between long-term aircraft noise exposure and heart disease or stroke, but there are very few such studies on short-term noise exposure.

We first looked at how aircraft noise varies across the day in areas affected by noise from aircraft arriving at and departing from London Heathrow airport. We used standard noise models that use information such as flight paths, type of aircraft and weather conditions to estimate aircraft noise levels at different times of day near Heathrow airport in 2014–18. We found that the daytime periods 7 a.m.–7 p.m., with the largest number of flights, had higher noise levels than evening or night-time and higher numbers of noisy flights. However, the early morning (6 a.m.–7 a.m. had the highest average noise levels. Night-time aircraft noise levels were lower but fluctuated more than at any other time of day.

We investigated inequalities in noise exposures by comparing wealthy with less wealthy areas and found that wealthier areas tended to have lower aircraft noise levels, especially at night.

We then examined whether higher noise levels at particular times of day in an area were linked with higher hospital admissions and deaths from heart disease or stroke (cardiovascular disease). We saw a small increased risk of hospital admissions for cardiovascular disease if there were high evening noise levels the previous day. This may be linked to sleep disturbance. Men aged over 65 years also showed increased risks associated with daytime aircraft noise.

Finally, we assessed whether areas prone to changing aircraft noise patterns (i.e. with relief periods from aircraft noise) affected the increased risk of cardiovascular disease in areas with higher noise in the evenings and found the higher risks were only seen in areas with more constant noise levels. More research is needed to investigate potential health benefits of relief periods with lower noise.

Scientific summary

Background

Long-term exposure to aircraft noise has been associated with small increases in risk for cardiovascular health outcomes but there are almost no short-term exposure studies. Relief periods are valued by residents under flight paths, whether in relation to planned flight path changes or because of weather patterns, but it is unclear whether these relief periods have population health benefits.

Research questions and objectives

The specific research questions were:

1. Is there a significant short-term impact of aircraft noise on cardiovascular morbidity and mortality?
2. Are there interactions with factors such as age, gender, ethnicity and deprivation that may function as effect modifiers?
3. Is there variability in risk estimates between areas with consistent patterns of noise exposure versus those with changing patterns of noise exposure?
4. How do risk estimates differ when using different noise metrics?

Objectives were:

1. to obtain daily estimates of day and night-time noise average exposure and number of noisy (flight) events for 2014–18 for the population living around London Heathrow airport;
2. to link the noise estimates to cardiovascular hospital admission and mortality data via postcode of residence;
3. to conduct a case-crossover analysis relating daily changes in aircraft noise to cardiovascular disease morbidity and mortality, taking account of relevant confounders that also change on a day-to-day basis such as temperature and air pollution;
4. to identify relevant interactions for areas with consistent patterns of noise exposure versus those with changing patterns of noise exposure and to explore interactions with age, gender, ethnicity and deprivation.

Methods and results

Data sources

- We worked with a specialist noise consultancy that modelled aircraft noise at different times of day, every day from 2014 to 2018, using a standard noise model, the Aviation Environmental Design Tool, for (1) noise levels expressed in decibels as equivalent continuous sound levels (LAeq) for eight time bands, and (2) number of aircraft noise events above defined noise thresholds (2018 only). The time bands were defined by the study scientific advisory board (relating to night-time, morning shoulder, morning, afternoon, evening, late evening, night-time shoulder, which correspond to aircraft operation periods). We provided input data for the model such as temperature and evaluated outputs against annual average Civil Aviation Authority (CAA) noise levels (that use the ANCON noise model).
- Our health outcomes were NHS Digital hospital admission records and mortality records from the Office for National Statistics for 2014–18 for cardiovascular outcomes, plus individual-level factors available from healthcare records (e.g. age and sex). We used data held by the Small Area Health Statistics Unit.

- We obtained confounder data from a number of sources including road traffic noise (University of Leicester modelled), rail noise (Department for Environment, Food and Rural Affairs), air pollution, area-level deprivation measures of Carstairs index (UK Census), avoidable death rate (UK government statistics), fuel poverty (Office for National Statistics) and ethnicity (NOMIS, provided by the Office for National Statistics).

Study area, study unit and population

The study area was designed to capture the outer bounds of the CAA annual-average aircraft noise contours in 2011 (these are produced around every 5 years; 2011 was also a Census year). This encompassed a boundary box that extends approximately 97 km east to west and 47 km north to south. Between 2014 and 2018, this included between 155,448 and 156,324 postcodes annually.

We used postcodes as the unit of analysis. On average, each postcode in the study area contained 53 residents [standard deviation (SD) 44] and 22 households (SD 17). In 2011, the total population of this boundary box was approximately 6.3 million people.

Descriptive analyses

The morning shoulder period (06.00–07.00 hours; mean: 50.92 dB; 90th percentile: 52.08 dB) and daytime (07.00–15.00 hours) were the noisiest periods (mean: 49.87; 90th percentile: 51.50 dB). On average, the night shoulder and night quota periods (23.30–04.30 hours) were the quietest.

Postcodes within the study area during daytime 07.00–15.00 hours experienced an average of 8 noisy flight events (> 65 dB), with 10% of postcodes experiencing 10 events. Morning shoulder (06.00–07.00 hours) had the second highest (90th percentile) but the third highest mean noisy flight events (mean = 3; 90th percentile = 9). During the night quota period (04.30–06.00 hours), the average number of noisy flight events (> 60 dB) per postcode was one.

Approaches to identifying respite and/or relief periods

We did not have information about trials or operational changes of flight paths over the period of the study. However, trials tend to cover relatively small areas. We identified one doctoral dissertation in the literature that used a natural experiment of the Early Morning Arrival Trial at Heathrow. The study examined the impact of night and early morning flight rerouting on medical expense, within four exclusion zones (two to the east and two to the west of Heathrow). During the trial (5 November 2012–31 March 2013), each week, the night and early morning (23.30–06.00 hours) aircraft movement was rerouted from one set of air traffic exclusion zones to non-exclusion areas. A difference-in-difference analysis found no difference in expenditure for CVD in zones where respite was implemented compared with control zones, but size of the affected area was small.

In our experiment, we therefore used the natural feature of wind direction changes that alter the flight paths as aircraft take off into the wind (broadly speaking 70% of the time Heathrow operates on a westerly pattern and 30% on an easterly), which meant that we could select from all days between 2014 and 2018 and all areas near the airport.

Our first approach was to identify areas that experienced pre-defined differences in noise exposures. Published laboratory research conducted for Heathrow airport (<https://www.heathrow.com/company/local-community/noise/making-heathrow-quieter/respite-research>; accessed 13 February 2023) has suggested that differences of 5–6 dB between successive sounds may be necessary for people to discern that there is a difference and that a difference of at least 7 or 8 dB may be needed between the average sound levels of two sequences of aircraft sounds to provide a valuable break from aircraft noise. We therefore investigated the number of areas that had detectable noise level changes compared with control areas that had more constant levels of noise using predefined cut-off levels referring to this report (e.g. 100+ days > 55 dB and 100+ days ≤ 50 dB in daytime periods). The number of postcodes

identified with 5 dB or greater noise differences and durations of noisy/quiet days was small. We did not proceed to conduct health analyses using these criteria because the small sample size may have been insufficient to detect effects.

The approach we explored for health analysis was empirical and based on noise variability as seen in the whole dataset. To identify areas with changes in noise exposures we calculated the coefficients of variation (CoV) of daily aircraft noise levels by postcode for the entire dataset (all four seasons) or by season (summer, summer transition, winter and winter transition). We found that night-time (24.00–04.30 hours) had the highest mean CoV (67.33–74.16), followed by 04.30–06.00 and 23.00–24.00 hours. The variability of daytime aircraft noise tended to be lower.

Investigating daily aircraft noise exposure and material and health inequality

There were inequalities in aircraft noise exposure. We employed a random effects model with autoregressive first-order autoregression model disturbance to investigate the relationship between noise levels and quintiles of deprivation. We explored relationships with three different area-level measures related to deprivation: the Carstairs index (a composite measure from UK Census variables relating to poverty), the avoidable death rate (health inequality) and fuel poverty (wealth inequality).

We found that postcodes near Heathrow airport within quintile 1 (least deprived) of either the Carstairs index or avoidable death rate experienced the lowest daily noise levels between 2014 and 2018. While there was no clear exposure–response relationship between two deprivation measures (Carstairs index and fuel poverty) and three noise metrics [equivalent continuous sound pressure levels during 07.00–19.00 hours (L_{day}), during 19.00–23.00 hours (L_{eve}) and for 24 hours], we observed a gradient between equivalent continuous sound pressure level during 23.00–07.00 hours (L_{night}) and Carstairs index and avoidable death rate.

Short-term impact of aircraft noise on cardiovascular disease

We used a time-stratified case-crossover study design, in which the days when an outcome of interest occurred are matched with control days within the same month and on the same day of the week and the exposure on case and control days are compared. This approach accounts by design for confounding variables that are invariant or slowly time variant, such as age and sex. We adjusted for confounding variables that change over short periods, such as concentrations of particulate matter less than 2.5 µm in diameter, temperature and holiday periods. We included all recorded hospitalisations ($n = 442,442$) and deaths ($n = 49,443$) in 2014–18 due to CVD in the analysis and used conditional logistic regression to estimate odds ratios (OR). We observed a statistically significant increase in risk for CVD hospital admissions for a 5-dB increment in noise during L_{eve} [Level OR 1.005, 95% confidence interval (CI) 1.000 to 1.010], particularly from 22.00 to 23.00 hours [OR 1.006, 95% CI 1.002 to 1.010], but did not detect statistically significant associations for other periods or for mortality. If the association is causal, it suggests that sleep disturbance may be a mechanism.

We found effect modification by age, sex, ethnicity, deprivation and season (winter, winter transition, summer and summer transition).

When stratified by CoV, our results showed a statistically significant adverse association between evening noise levels (19.00–22.00 hours, 22.00–23.00 hours and 23.00–24.00 hours) and hospital admission for CVD in low (below mean) CoV postcodes but not in high CoV postcodes. To explore whether fewer relief periods these low CoV areas were those with higher noise levels (potentially suggesting high noise and less relief periods), we examined mean noise levels. For the latter two periods, mean noise levels were higher in the low CoV postcodes (41 dB vs. 37 dB for 22.00–23.00 hours; 31 dB vs. 24 dB for 23.00–24.00 hours). However, for the period 19.00–22.00 hours, the mean noise levels were 41 dB in low CoV areas compared with 43 dB in high CoV areas. We therefore cannot readily infer that lack of relief periods (or at least some periods of lower noise exposure) was associated with hospitalisation.

Conclusions

Our study focused on the impact of short-term aircraft noise on cardiovascular morbidity and mortality. Our findings suggest an association between short-term exposure to noise during evening and night-time hours, and an elevated risk of hospital admissions (but not deaths) for CVD. Our findings also suggested that the variability of noise exposure may play an important role in its relationship with health outcomes. Our results could be useful for residents, future health studies and a variety of other studies.

Recommendations for future research

- Further studies are needed to assess the impact of intervention on short-term aircraft noise exposure on CVD – this is one of the first such studies to date.
- Further research is needed to investigate the relationship between noise variability and the risk of CVD, potentially starting with laboratory experimental studies or field studies of flight path changes, and using intermediate continuous outcomes, such as blood pressure, rather than binary outcomes.
- Further research is needed to explore effect modifiers, such as introducing noise insulation measures for areas most affected by aircraft noise, which may have important implications for future policy interventions.
- More research is needed to explore associations between deprivation and noise exposure levels.
- Exploring the relationship between outdoor and indoor noise exposure levels is important.

Implications for future studies

- Standard noise metrics such as L_{day} and L_{night} may not capture periods where exposure has most impact on health and use of alternative noise metrics need to be explored.
- Future studies on the health effects of aircraft noise pollution are advised to take noise variability into account.
- Future epidemiological studies are recommended to consider different metrics of health deprivation as a potential confounder and effect modifier.

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Chapter 1 Background

In areas affected by aircraft noise near major airports, relief periods of lower or absent aircraft noise are valued by residents. These periods may occur in relation to deliberate flight path changes to provide respite or because of changes due to weather variations, especially wind direction. However, it is unclear whether this respite might result in benefits to health endpoints. There is certainly good evidence that long-term exposure to aircraft noise can affect quality of life, cause annoyance and disturb sleep.¹ In recent years, evidence has accumulated which suggests that long-term exposure to transport noise may result in high blood pressure and may also impact cardiovascular health. A meta-analysis of 24 studies of road traffic noise² published in 2012 and two meta-analyses each of five studies of aircraft noise^{3,4} published in 2009 and 2015 (comprising seven studies in total) found associations between long-term aircraft noise exposure and hypertension in adults. A meta-analysis from Vienneau *et al.* in 2015 included 10 studies on transport noise⁵ (3 of which included aircraft noise) and found associations between long-term noise exposure and ischaemic heart disease (IHD). Vienneau *et al.*'s updated meta-analysis of 13 studies (13 estimates for road traffic, 5 for aircraft and 3 for railway noise exposure) in 2019 found a 2%, 3% and 1% increase in relative risk of IHD per 10 dB increment in day-evening-night noise levels (Lden) for road, aircraft and railway noise exposure, respectively (although only risk for road traffic noise was statistically significant).⁶ A meta-analysis conducted by some of the applicants and authors of this report, which included relevant studies from two meta-analyses^{6,7} and one literature review,⁸ as well as new studies published until February 2022, found a 3% and 2% increased risk for IHD incidence and mortality per 10 dB Lden for aircraft noise, respectively, but the risk for mortality was not statistically significant.⁹ The applicants' own *BMJ* study published in 2013¹⁰ found that higher long-term average aircraft noise exposures in areas near London Heathrow airport were associated with higher average risks of hospital admission and mortality for heart disease and stroke.

The mechanisms for such effects may include impacts on the autonomic nervous system, either directly or indirectly via psychological annoyance that results in a stress reaction, and/or lack of restorative sleep, both of which will impact on cardiovascular health.¹¹ Although exposure to aircraft noise could induce adaptation in humans, there is evidence that adaptation to loud noise is typically incomplete,¹² in part due to diminished coping resources, as aviation noise sources are typically outside the control of the individual.¹³

Extremely few studies have examined the impact of short-term changes in transport noise on health. Recent experimental studies in humans have documented short-term rises in blood pressure and cardiovascular risk markers in the blood following aircraft noise exposure during sleep¹⁴ and a small panel study has shown changes in heart rate variability can be induced with daytime noise exposures (not specifically transport noise).^{15,16} A case-crossover study of cardiovascular mortality in relation to road traffic noise in Madrid over a 3-year period,^{15,16} recently updated to 7 years of follow-up,¹⁷ found short-term (lag 0 and lag 1) increases in IHD and myocardial infarction mortality that were independent of air pollution exposures. This road noise study may not be directly applicable to aircraft noise, which is a qualitatively different type of sound. Also, aircraft noise may vary around airports due to changes in flight paths because of wind direction and operational conditions, which is not the case for roads, which have fixed geographical positions.

We are aware of only one previous study on short-term impacts of aircraft noise on cardiovascular outcomes. This is a case-crossover study by Saucy (2021) examining around 25,000 cardiovascular deaths near Zürich airport, Switzerland, taking data from the Swiss national cohort. Using the fact that the Swiss mortality records have precise time of death, the authors found that aircraft noise exposure levels in the 2 hours preceding death for night-time deaths were significantly associated with cardiovascular disease (CVD) mortality [odds ratio (OR) 1.44, 95% confidence interval (CI) 1.03 to 2.04] for the highest exposure group [equivalent continuous sound pressure level (LAeq) > 50 dB vs. < 20 dB].

BACKGROUND

Aircraft noise was assigned to place of residence; one limitation of the study is that it was not clear whether this was also place of death.

To our knowledge, there is only one study that has examined short-term flight changes in aircraft noise on hospital admissions – the closure of Heathrow airport in 2010 for 6 days following eruption of Iceland's Eyjafjallajökull volcano.¹⁸ The authors used an interrupted time series design but were unable to detect changes in CVD hospital admissions in areas within the 55 dB(A) noise contour of Heathrow (0.7 million population) over this 6-day period. The authors comment that this may be related to lack of statistical power. To give confidence that analyses can actually detect what might be small increases in risk, such studies would need to include large numbers of people and detailed long-running information on daily aircraft noise levels. Daily variability in population noise exposure is not available from standard sources such as the Civil Aviation Authority (CAA), which is addressed in the proposed research.

Chapter 2 Objectives

The aim of this study was to evaluate the potential short-term impact of aircraft noise exposure on cardiovascular morbidity and mortality in a general population. This is one of the first studies to examine short-term associations of aircraft noise with cardiovascular outcomes and, additionally, to consider impacts of changes in noise levels. The results have potential inference for interventions that reduce aircraft noise levels reduce CVD outcomes in the short term.

This study used variability in night and daytime aircraft noise related to operational and weather-related (e.g. easterly/westerly wind) changes in flight paths in the area around London Heathrow airport, which is one of the top 10 busiest airports in the world, sited in close proximity to a densely populated urban area. Our previous study¹⁰ found that around 3.6 million people live around London Heathrow airport, who are potentially affected by aircraft noise. There is intense policy interest in studies around London Heathrow, given the potential expansion and addition of a third runway currently under consideration. Given the number of local residents affected and high public interest within London in aircraft noise, Heathrow has one of the most stringent approaches to noise control of major world airports and has conducted a number of trials of flight paths to try to improve noise exposures of local residents. Results from this study should be transferable to other countries.

The specific research questions that we attempted to answer are:

1. Is there a significant short-term impact of aircraft noise on cardiovascular morbidity and mortality?
2. Are there interactions with factors such as age, gender, ethnicity and deprivation that may function as effect modifiers?
3. Is there variability in risk estimates between areas with consistent patterns of noise exposure compared with those with changing patterns of noise exposure?
4. How do risk estimates differ when using different noise metrics?

The objectives were:

1. To obtain daily estimates of day and night-time noise average exposure and the number of noisy events for 2011–15 for the population living around London Heathrow airport.
2. To link the noise estimates to cardiovascular hospital admission and mortality data via postcode of residence.
3. To conduct a case-crossover analysis relating daily changes in aircraft noise to CVD morbidity and mortality, accounting for relevant confounders that also change on a day-to-day basis, such as temperature and air pollution.
4. To identify relevant interactions for areas with consistent patterns of noise exposure versus those with changing patterns of noise exposure, and to further explore interactions with age, gender, ethnicity and deprivation.

Chapter 3 Methods – generating highly time–space resolved aircraft noise exposure data

Figure 1 presents the details of the research pathway, indicating which institutions did which tasks.

We began by introducing the study area, population and period. The model and input data used to generate daily aircraft noise exposure data were then discussed.

Study area, unit and population

The study area's centroid was at Heathrow airport, covering a bounding box with longitudes extending from -0.901° to -0.031° west, and latitudes from 51.345° to 15.609° north. It covered an approximate distance of 97 km east to west, and 47 km north to south. The study area was designed to capture the outer bounds of the CAA annual average aircraft noise contours in 2011 (Figure 2) that were available when the study was being designed and could be aligned with the population data from the UK 2011 Census (the UK Census takes place every 10 years).

To reduce the computational demands of modelling, a grid resolution of 100×100 m was specified near to Heathrow, with a 200×200 -m resolution then used to the extent of the study area. The inner grid, with a 100-m resolution, covered the area from Datchet to Osterley Park (approximately 25 km east to west) and West Drayton to Ashford (approximately 15 km north to south).

We used postcodes as the unit of analysis because they represented the smallest geographical area in the UK, allowing us to model noise levels with the highest possible spatial resolution (approximately 1.75 million live postcodes across the country in 2016¹⁹). Postcodes are designed to support postal mail

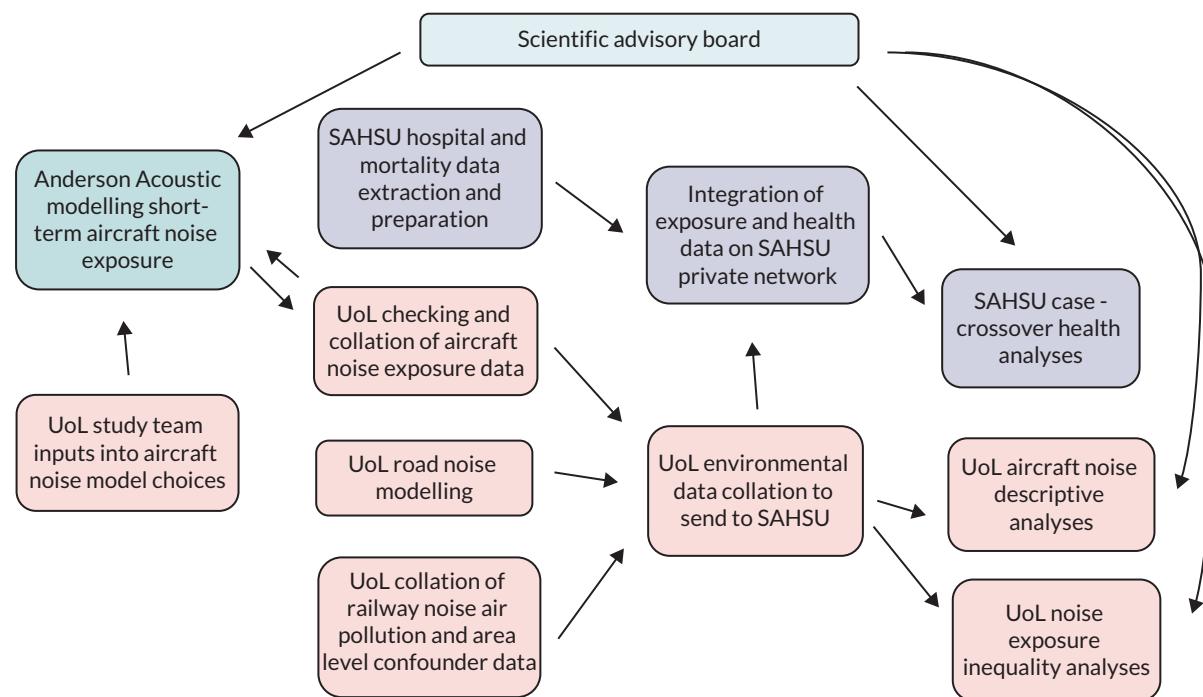
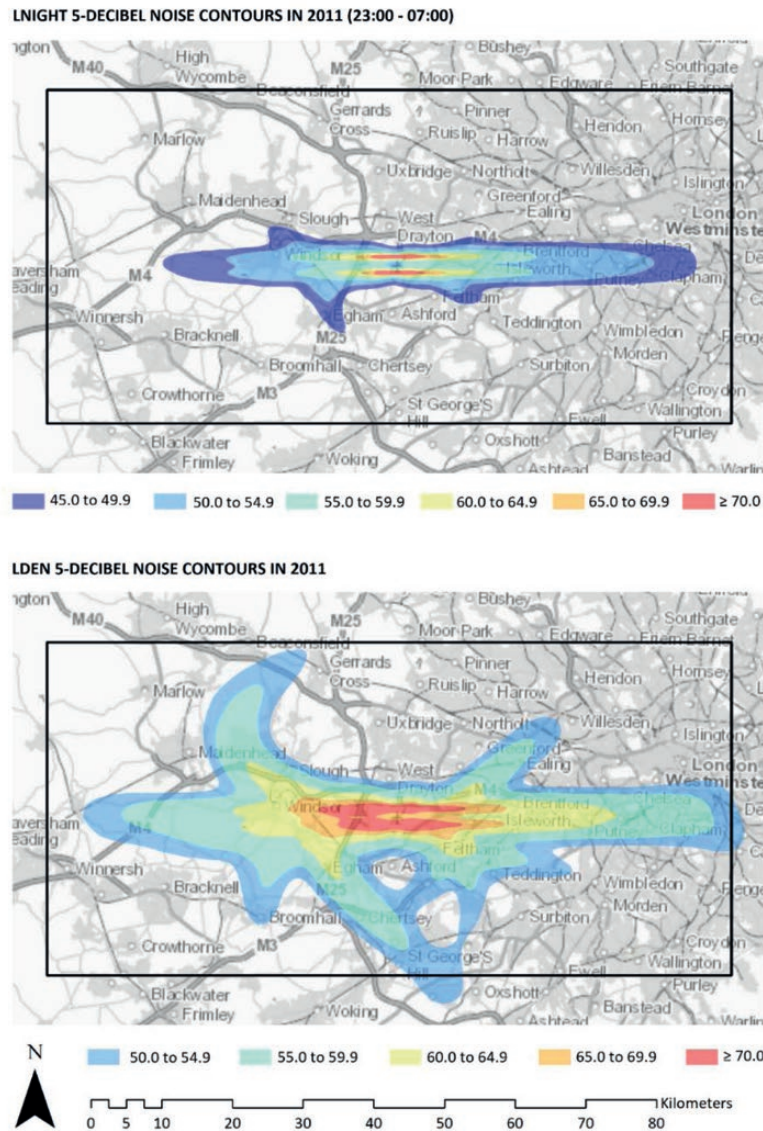


FIGURE 1 Research pathway diagram. Key: orange boxes – work conducted at University of Leicester (UoL); purple boxes – work conducted at Small Area Health Statistics Unit (SAHSU), Imperial College, London; cyan box – work conducted by noise consultancy.



(a) Lnight and Lden noise contour

FIGURE 2 The spatial extent of the Aviation Environmental Design Tool modelling exercise (black bounding box) in relation to the CAA annual average aircraft noise contours for 2011 for Lnight (top) and Lden (bottom). (a) Lnight and Lden noise contour. (b) Lnight and Lden noise contour (base map 50% transparency). (c) Lnight and Lden noise contour (base map 60% transparency). Note: (a) Full Lnight and Lden noise contour, (b) and (c) are the same contour maps but magnified while maintaining a background map with a 50% and 60% transparency level, respectively. (continued)

deliveries and small numbers of postcodes in the study area change every year, as new postcodes are created, and old postcodes become redundant from increases in and redistributions of the population over time. The total numbers of postcodes in each year are listed in [Table 1](#).

Typically, each postcode within the study area consists of 53 residents [standard deviation (SD) 44] and 22 occupied households (SD 17), based on headcount data from Nomis.²⁰ The combined population of this boundary box in 2011 was approximately 6.3 million.

Model

Version 3b of the Aviation Environmental Design Tool (AEDT), developed by the US Federal Aviation Administration, was used to assess aircraft noise levels at each of the postcodes within the study area.

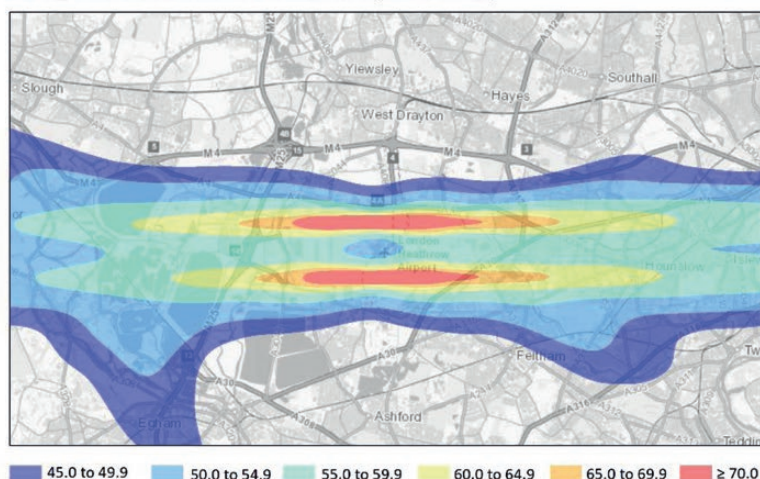
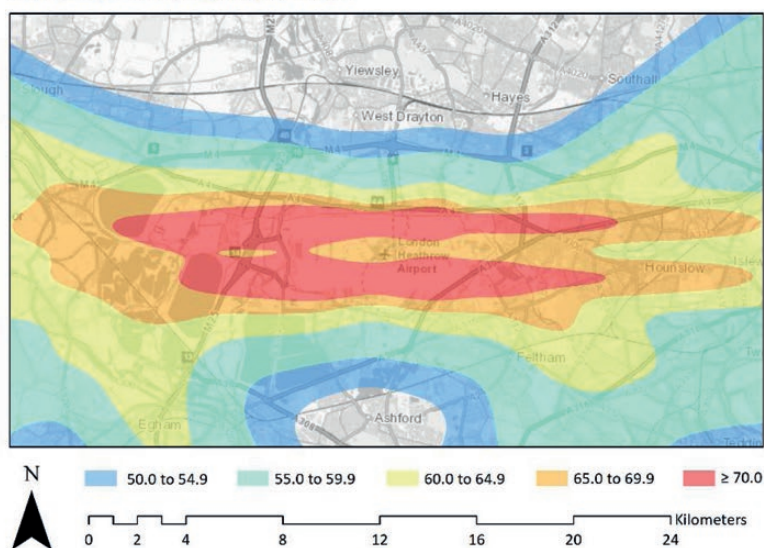
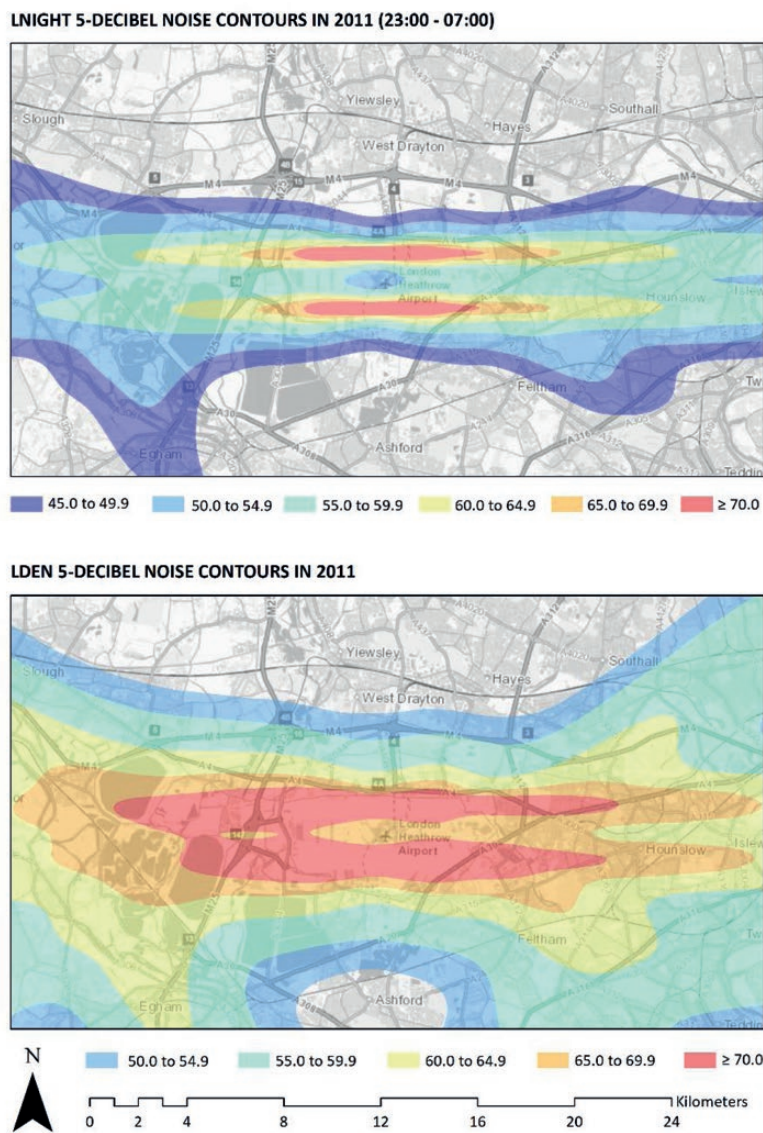
NIGHT 5-DECIBEL NOISE CONTOURS IN 2011 (23:00 - 07:00)**Lden 5-DECIBEL NOISE CONTOURS IN 2011****(b) Nlight and Lden noise contour (base map 50% transparency)**

FIGURE 2 The spatial extent of the Aviation Environmental Design Tool modelling exercise (black bounding box) in relation to the CAA annual average aircraft noise contours for 2011 for Nlight (top) and Lden (bottom). (a) Nlight and Lden noise contour. (b) Nlight and Lden noise contour (base map 50% transparency). (c) Nlight and Lden noise contour (base map 60% transparency). Note: (a) Full Nlight and Lden noise contour, (b) and (c) are the same contour maps but magnified while maintaining a background map with a 50% and 60% transparency level, respectively. (*continued*)

The AEDT was developed to model aircraft performance in space and time to estimate noise, fuel consumption, emissions and air quality consequences.²¹ This tool is actively used by the US government for regulatory studies, research and domestic aviation system planning, as well as domestic and international aviation environmental policy analysis.

Local parameters

Operational procedures and fleet profiles for Heathrow were extracted from the AEDT database. This includes the geographical content for activity around the surface structures (i.e. runways, taxiways and terminals) and airspace (i.e. ground tracks, altitude controls, etc.). Fleet profiles are also provided by (the



(c) Lnight and Lden noise contour (base map 60% transparency)

FIGURE 2 The spatial extent of the Aviation Environmental Design Tool modelling exercise (black bounding box) in relation to the CAA annual average aircraft noise contours for 2011 for Lnight (top) and Lden (bottom). (a) Lnight and Lden noise contour. (b) Lnight and Lden noise contour (base map 50% transparency). (c) Lnight and Lden noise contour (base map 60% transparency). Note: (a) Full Lnight and Lden noise contour, (b) and (c) are the same contour maps but magnified while maintaining a background map with a 50% and 60% transparency level, respectively.

TABLE 1 The number of postcodes in existence in the years 2014–18

Year	Postcodes
2014	156,324
2015	155,960
2016	155,558
2017	155,448
2018	155,671

European Organisation for the Safety of Air Navigation) EUROCONTROL family and International Civil Aviation Organization engine performance.

Heathrow airport's radar system provided records of flight activity, which included the position, height and speed of all aircraft for 2014–18. The headwind direction was determined by the actual direction of operation. The headwind speed was maintained at the AEDT default of 8 knots during the entire period of each operation. That is, wind direction changes such that it is always directed against aircraft course.

Wind measurements are often used to approximate the headwind direction and aircraft performance parameters such climb (therefore height) and speed; however, such information was comprehensively captured by the radar data system at Heathrow. Wind speed or direction is not used by the AEDT sound propagation calculations, which may be viewed as a limitation of current modelling practices.

Several meteorological parameters were included in the AEDT noise calculations:

- We evaluated the dry air temperature data measured at Heathrow airport, to reduce the number of required models run. Fluctuations in average temperature were evaluated for 2014–18 by month and time of day (Tables 2–4). The hourly mean dry air temperature measurements were summarised into 32 groups, based on season ($n = 4$) and time of day ($n = 8$).
 - Annual mean temperatures are relatively stable for the period 2014–18, ranging from 1.6 to 12.4°C.
 - Seasonal variations in temperature exist. Mean winter months temperatures at Heathrow are 5.8–9.0°C (November–March) and mean summer month temperatures are 14.0–19.5°C (June–August). Summer month temperatures are generally more stable, with lower levels of relative

TABLE 2 Hourly dry air temperature (°C) profiles summarised into monthly and annual means with their coefficient of variation^a

Time period		Count (N)	Mean (°C)	SD (°C)	CoV (%)
All observations	1 January 2014 to 31 December 2018	43,080	12	6	49.9
Month	January	3720	5.9	3.5	58.9
	February	3378	5.8	3.5	59.1
	March	3716	7.9	3.7	46.6
	April	3095	10.5	3.8	36
	May	3715	14	4.1	29.4
	June	3600	17.3	4.1	23.6
	July	3710	19.5	4.2	21.4
	August	3715	17.9	3.8	21.2
	September	3542	15.6	3.7	24
	October	3718	12.6	3.4	26.7
	November	3451	9	3.6	40.4
	December	3720	7.7	3.9	51.3
Year	2014	8752	12.4	5.5	44.5
	2015	8256	11.9	5.4	45.5
	2016	8773	11.6	6.1	52.3
	2017	8758	12	6.1	50.6
	2018	8541	12.3	6.8	55.5

^a The CoV is a way to quantify scatter. It is defined as the SD of a group of values divided by their mean.

TABLE 3 Hourly mean dry air temperature (°C) at Heathrow airport for 2014–18 by month

	00.00	01.00	02.00	03.00	04.00	05.00	06.00	07.00	08.00	09.00	10.00	11.00	12.00	13.00	14.00	15.00	16.00	17.00	18.00	19.00	20.00	21.00	22.00	23.00
January	5.1	5	4.9	4.8	4.7	4.6	4.7	4.8	4.9	5.5	6.3	7	7.5	7.8	7.8	7.6	7	6.6	6.3	6.1	5.9	5.8	5.6	5.4
February	4.9	4.7	4.5	4.4	4.3	4.3	4.3	4.4	4.8	5.7	6.5	7.2	7.6	7.8	8	7.8	7.5	7	6.5	6.2	5.8	5.6	5.3	5.1
March	6.3	6	5.8	5.6	5.5	5.4	5.4	5.9	6.8	7.9	8.8	9.7	10.4	10.7	11	10.9	10.5	9.9	9.1	8.5	8	7.5	7.1	6.7
April	8.3	7.9	7.6	7.4	7.2	7.2	7.7	8.8	10	11.2	12.1	12.9	13.4	13.8	14	13.9	13.6	13.1	12.3	11.4	10.7	10	9.3	8.7
May	11.3	10.9	10.6	10.3	10.2	10.5	11.5	12.7	13.8	14.9	15.8	16.5	17.1	17.4	17.6	17.5	17.3	16.8	16.1	15.1	14.2	13.4	12.6	12
June	14.5	14	13.6	13.3	13.2	13.7	14.6	15.7	16.9	17.9	18.9	19.7	20.4	20.9	21.2	21.2	20.9	20.5	19.8	18.8	17.7	16.8	15.9	15.1
July	16.9	16.3	15.9	15.5	15.4	15.7	16.7	17.8	19	20.2	21.2	22	22.6	23.1	23.4	23.4	23.1	22.7	22.1	21.2	20.1	19.1	18.2	17.5
August	15.6	15.2	14.8	14.6	14.3	14.4	15.2	16.3	17.5	18.7	19.7	20.4	20.9	21.3	21.5	21.5	21.2	20.7	19.9	18.8	18	17.3	16.6	16
September	13.4	13.1	12.8	12.6	12.4	12.4	12.8	13.6	14.9	16.3	17.4	18.2	18.8	19.2	19.3	19.1	18.7	18	17	16.2	15.5	14.9	14.3	13.8
October	11.1	10.8	10.7	10.6	10.6	10.6	10.7	11.1	11.9	13	13.9	14.7	15.3	15.5	15.6	15.3	14.7	13.9	13.3	12.8	12.5	12.1	11.7	11.3
November	8	7.9	7.7	7.6	7.5	7.6	7.7	7.8	8.2	9	9.8	10.4	10.9	11.1	11.1	10.7	10.2	9.8	9.5	9.2	8.9	8.7	8.4	8.1
December	6.9	6.7	6.6	6.6	6.6	6.7	6.7	6.8	6.9	7.4	8.1	8.7	9.2	9.4	9.4	9.1	8.6	8.3	8	7.9	7.7	7.6	7.4	7.2

Note

The conditional format style is based on a three-colour scale, with the lowest (min) value set as red, the highest value (max) set as red, and the midpoint set as white is defined as the 50th percentile (med). This visually identifies where individual hours fit within the range of temperature values across the course of a typical day.

TABLE 4 Mean dry air temperature (°C) values at Heathrow airport, for the irregularly constructed periods of time (2014–18)

Group	Months	07.00–15.00	15.00–19.00	19.00–22.00	22.00–23.00	23.00–24.00	24.00–04.30	04.30–06.00	06.00–07.00
Winter	November to March	8.0	8.5	7.3	6.8	6.5	5.9	5.7	5.8
Winter transition	April and October	13.0	13.8	11.6	10.5	10.0	9.2	8.9	9.2
Summer transition	May and September	16.5	17.6	14.9	13.5	12.9	11.7	11.4	12.2
Summer	June to August	19.9	21.4	18.6	16.9	16.2	14.8	14.5	15.5

dispersion around the mean. For each model run, the AEDT model used one of the 32 unique temperature values from the 5-year profiles shown in [Table 4](#), which corresponds to the specified 'day' and 'time of day'. In total, 1826 days across the 5 years of 2014–18 were considered, with each day comprising eight irregularly grouped periods (i.e. 14,608 flight-activity noise model surfaces were created).

The Ordnance Survey digital terrain model of Great Britain, which has a 50-m horizontal resolution, was passed into the AEDT model to create terrain features. It is an open-height dataset of contours with spot heights, break lines, coastline, lakes, ridges and form lines cross Great Britain with a typical accuracy level greater than 2 m root mean square deviation. With terrain elevation processing, it is possible to adjust sound propagation from the attenuation due to line of sight blockage. It is based on the difference in propagation path length between the direct path and propagation path over the top of terrain features, known as path length difference. The terrain model used only accounts for the elevation of natural landscapes and not manmade features (i.e. buildings).

Model output

A comprehensive set of average 'A' frequency weighted noise estimates were provided for eight time bands (see [Table 3](#)) over the course of each day (LAeq) for the study period of 1 January 2014 to 31 December 2018. These periods were chosen in discussion with the study advisory board, including industry representatives, to capture conventional periods (i.e. 07.00–19.00 hours – day, 19.00–23.00 hours – evening, 23.00–07.00 hours – night), together with timings that are aligned with Heathrow operations (e.g. 23.30–04.30 hours scheduled night flight ban, 07.00–15.00 hours and 15.00–22.00 hours respite periods). The 'A' weighting is the standard weighting of the audible frequencies designed to reflect the response of the human ear to noise (between 500 Hz and 6 kHz).

We used daily noise levels during eight period bands, as mentioned above, to further calculate daily mean aircraft noise levels in four metrics: Lday (07.00–19.00 hours), Leve (19.00–23.00 hours), Lnight (23.00–07.00 hours) and LAeq24 (24-hour average).

In addition, the daily number of flight events exceeding a maximum sound level of 65 A-weighted decibels [dB(A)] in the daytime and 60 dB(A) at night, were estimated at each modelled location (N-Above) from 1 January 2018 to 31 December 2018. For further context, if any of the one-eighth-second periods from an aircraft noise event (generated by a single aircraft operation) exceeds the specified threshold, then that event is counted as one. The N-Above measure cannot exceed the number

of operations that occur in the specified period. These event counts are based on the maximum sound level with 'A' frequency weighting and fast time weighting.

Limitations identified in the Aviation Environmental Design Tool model

We identified several limitations in the use of the AEDT model, particularly as we extended its use to look at short periods within a single day (it is usually used to provide long-term average noise exposures).

1. Atmospheric pressure, relative humidity and wind speed are set as meteorological constants that reflect the 30-year average at the airport. These simplifications are a limit of current modelling practices, when estimating sub-annual average aircraft noise exposures.
2. The headwind speed is maintained at 8 knots during the entire period of each operation. This may result in inaccurate aircraft performance parameters, such as climb and speed, which are related to the location and intensity of noise.
3. Wind speed or direction is not used by the AEDT sound propagation calculations (i.e. a uniform dispersion in all directions is assumed at all times).
4. The terrain model only accounts for elevation of natural landscapes and not manmade features. Therefore, containment and sheltering effects in urban locations are ignored.
5. The computational demands for creating sub-daily exposure surfaces:
 - a. Limited the spatial resolution of the model outputs, returning a coarser exposure gradient, although we think that this still gave good exposure contrast for our epidemiological study.
 - b. Drier air temperatures were summarised into profiles that accounted for season and time of day across the 5-year study period. Therefore, the influence of unusual temperature events on sound propagation is not accounted for.

These factors are likely to lead to exposure misclassification bias. However, annual average aircraft noise surfaces are currently only routinely modelled by the CAA. This study has used several approaches to develop and enhance the existing approach to create sub-annual exposure surfaces:

1. Radar tracks of individual flights were provided by Heathrow airport, with a unique set of aircraft footprints constructed for each modelled period.
 - A. The created AEDT surfaces cover 1826 days across the 5 years of 2014–18 (i.e. 14,608 flight activity-informed noise surfaces were created vs. 5 annual average surfaces for each noise metric).
 - B. Actual flight paths were used rather than these being estimated by operational movements being estimated by headwind direction and performance parameters.
2. Unique temperature profiles were used, which correspond to the specified 'season' and 'time of day'. Annual average AEDT models only use long-term averages.

We also had a very large dataset to work with, which may offset some of the lack of precision as a result of random bias.

Methods and results: descriptive analysis of daily aircraft noise data

We first examined the descriptive summary of the noise data. The study area annually had between 155,448 and 156,324 postcodes for the period 2014–18. We calculated log₁₀ logarithmic means, SDs and the 90th percentile for noise levels during the eight time bands (04.30–06.00, 06.00–07.00, 07.00–15.00, 15.00–19.00, 19.00–22.00, 22.00–23.00, 23.00–24.00, 24.00–04.30 hours), as well as four metrics, including LAeq24, Lday (07.00–19.00), Leve (19.00–23.00) and Lnight (23.00–07.00). These statistics were then anti-log transformed and are presented in [Table 5](#). We also presented arithmetic means, SDs and the 90th percentile of number of flight events per time band in the same table.

The number of noise observations per period varied throughout the day, with missing values due to the absence of air traffic above the corresponding postcodes. Four periods (06.00–07.00 hours, 07.00–15.00 hours, 15.00–19.00 hours and 22.00–24.00 hours) of the eight specified periods in this study had full numbers of observations ($N = 284,476,323$). Early morning (04.30–06.00 hours: $N = 283,706,122$) and night between 19.00 and 24.00 hours (19.00–23.00 hours: $N = 271,590,174$; 23.00–24.00 hours: $N = 279,444,325$) had slightly fewer number of observations, whereas night quota period (24.00–04.30 hours) had less than 31% of the total number of observations ($N = 87,705,638$). For standard noise metrics, Lday and Leve had 284,165,204 and 271,590,174 observations, respectively, which is close to the total number of postcodes. Comparatively, Lnight and LAeq24 had 86,618,974 and 83,220,954 observations, respectively, or approximately 29–30% of total number of postcodes.

The use of multiple periods throughout the day allowed us to explore which periods were the quietest and noisiest, looking at descriptive statistics for postcodes with values assigned (see [Table 5](#)). The morning shoulder period (06.00–07.00 hours) was the noisiest, with the highest mean (50.92 dB) and 90th percentile (52.93 dB) noise levels. Daytime (07.00–15.00 hours) aircraft noise levels had a mean noise level of 49.87 dB and the 90th percentile was 51.50 dB, the second highest of all periods. As the noise levels in each postcode in the study area exceeded 22.96 dB, aircraft noise affects nearly every postcode in the area. The quietest periods on average were night shoulder and night quota periods – the mean noise levels across postcodes during 23.00–24.00 hours and 24.00–04.30 hours were 41.06 and

TABLE 5 Descriptive summary of daily aircraft noise levels

Noise metrics	N	Mean	SD	Min	Max	P90
LAeq 04.30–06.00 hours	283,706,122	43.75	53.22	0	77.69	44.72
LAeq 06.00–07.00 hours	284,476,323	50.92	58.44	7.04	80.33	52.93
LAeq 07.00–15.00 hours	284,476,323	49.87	58.06	22.96	78.83	51.5
LAeq 15.00–19.00 hours	284,165,204	49.44	57.67	19.84	78.9	51.09
LAeq 19.00–22.00 hours	271,590,174	49.12	57.3	17.04	78.84	50.95
LAeq 22.00–23.00 hours	284,476,323	47.48	56.69	9.19	81.07	49.24
LAeq 23.00–24.00 hours	279,444,325	41.06	51.54	0	79.52	42.15
LAeq 24.00–04.30 hours	87,705,638	29.81	42.3	0	76.34	30.04
Lday	284,165,204	49.73	57.17	22.79	78.29	51.76
Leve	271,590,174	48.8	56.97	16.33	78.86	51.2
Lnight	86,618,974	44.19	51.39	4.23	74.13	46.49
LAeq24	83,220,954	48.92	55.98	22.87	76.88	50.72
N60 04.30–06.00 hours	56,819,915	1	2.62	0	33	1
N60 06.00–07.00 hours	56,819,915	3	7.39	0	58	9
N65 07.00–15.00 hours	56,664,244	8	32.91	0	388	10
N65 15.00–19.00 hours	56,819,915	4	16.47	0	199	5
N65 19.00–22.00 hours	56,819,915	3	11.87	0	146	3
N65 22.00–23.00 hours	56,819,915	1	2.83	0	47	1
N60 23.00–24.00 hours	56,352,902	0	1.34	0	43	1
N60 24.00–04.30 hours	21,171,256	0	0.67	0	23	1

P90, 90th percentile.

29.81 dB, respectively. The 90th percentiles were 42.15 and 30.04 between 23.00 and 24.00 hours and 24.00 and 04.00 hours. The descriptive summary of noise levels by season is presented in [Appendix 1](#).

The average Lday, Leve, Lnight and LAeq24 levels were 49.73, 48.80, 44.19 and 48.95 dB, respectively. Their 90th percentiles were 51.76, 51.20, 44.19 and 50.57 dB.

For noisy flight event numbers (N65: number of flights > 65 dB day; N60: number of flights > 60 dB night), the means, minima, maxima and 90th percentiles were rounded to the nearest integer to aid the interpretation of the results. The highest number of events occurred during 07.00–15.00 hours (see [Table 5](#)), with an average of 8 noisy flight events and with the top 10% of postcodes experiencing 10 events. Morning shoulder (06.00–07.00 hours) had the second highest 90th percentile, with a value of nine noisy flights, but the third highest mean, with a value of 3. During the night quota period (04.30–06.00 hours), the average number of flight events per postcode was one. Comparatively, 23.00–24.00 hours and 24.00–04.30 hours had an average of zero events.

For health analysis purposes, negative noise values as produced by models were ignored.

Correlations

In [Table 6](#), we present the pairwise Pearson correlation coefficients (r) between daily aircraft noise levels at eight specified time bands and standard four noise metrics (LAeq24, Lday, Lnight and Leve), and the daily number of flight events at eight time bands.

We found high to very high correlations (coefficient 0.68–0.90) between each pair of the four standard noise metrics (Lday, Lnight, Leve and LAeq24).

In comparison, daily aircraft noise during early morning (04.30–06.00 hours) and late night (24.00–04.30 hours), had a much weaker correlation (coefficient < 0.4) with noise levels during any other time bands. There were moderate to high correlations (coefficient 0.52–0.87) between each pair of daily aircraft noise levels during the day (07.00–15.00 hours), the afternoon (15.00–19.00 hours) and the early evening (19.00–22.00 and 22.00–23.00 hours).

There were weak to moderate correlations (coefficient 0.07–0.48) between the number of flight events and the actual noise levels during the eight time bands, with the exception of N60 06.00–07.00 hours, which had a moderate correlation with noise levels during 06.00–07.00, 07.00–15.00, 15.00–19.00 and 19.00–22.00 hours. Moreover, there were relatively weak correlations between each pair of noisy flight events except for N60 06.00–07.00 hours and N60 04.30–06.00 hours (coefficient 0.73), N65 19.00–22.00 hours and N65 15.00–19.00 hours (coefficient 0.96), N65 22.00–23.00 hours and N65 15.00–19.00 hours (coefficient 0.71), N65 22.00–23.00 hours and N65 22.00–23.00 hours (coefficient 0.55). This may show that the distribution of noisy flight events may differ from that of daily noise levels in dB.

The lower correlations of non-standard and noisy event metrics with the standard noise metrics of Lday, Lnight, Leve and LAeq24 (compared with high correlations between the standard metrics) raises the possibility that standard metrics may miss important characteristics of noise exposure, with potential relevance for impacts on biological systems.

Methods and results: approaches to identifying respite and/or relief period

An important question that we sought to answer was whether there was variability in cardiovascular health risk estimates between areas with consistent noise exposure patterns and those with changing noise exposure patterns. Changing noise exposure can be either a relief, defined as a break from or a

TABLE 6 Pairwise correlations between noise metrics

Variables	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)	(13)	(14)	(15)	(16)	(17)	(18)	(19)	(20)
(1) LAeq 04.30–06.00 hours	1.000																			
(2) LAeq 06.00–07.00 hours	0.680	1.000																		
(3) LAeq 07.00–15.00 hours	0.425	0.725	1.000																	
(4) LAeq 15.00–19.00 hours	0.343	0.632	0.845	1.000																
(5) LAeq 19.00–22.00 hours	0.309	0.574	0.764	0.868	1.000															
(6) LAeq 22.00–23.00 hours	0.090	0.299	0.520	0.562	0.664	1.000														
(7) LAeq 23.00–24.00 hours	–0.062	0.109	0.269	0.279	0.353	0.600	1.000													
(8) LAeq 24.00–04.30 hours	0.075	0.157	0.276	0.262	0.283	0.373	0.357	1.000												
(9) Lday	0.415	0.723	0.974	0.928	0.831	0.555	0.286	0.284	1.000											
(10) Leve	0.275	0.546	0.755	0.855	0.977	0.771	0.417	0.315	0.820	1.000										
(11) LAeq24	0.445	0.744	0.950	0.924	0.901	0.665	0.366	0.306	0.980	0.901	1.000									
(12) Lnight	0.682	0.912	0.769	0.698	0.682	0.495	0.310	0.289	0.778	0.680	0.818	1.000								
(13) N60 04.30–06.00 hours	0.296	0.440	0.421	0.377	0.385	0.229	0.066	0.114	0.431	0.369	0.465	0.497	1.000							
(14) N60 06.00–07.00 hours	0.339	0.620	0.596	0.519	0.513	0.309	0.113	0.163	0.601	0.496	0.631	0.661	0.733	1.000						
(15) N65 07.00–15.00 hours	0.190	0.344	0.519	0.390	0.370	0.328	0.206	0.209	0.503	0.393	0.500	0.413	0.331	0.497	1.000					
(16) N65 15.00–19.00 hours	0.176	0.308	0.390	0.484	0.400	0.410	0.229	0.201	0.449	0.433	0.477	0.394	0.293	0.400	0.433	1.000				
(17) N65 19.00–22.00 hours	0.169	0.295	0.373	0.462	0.403	0.431	0.243	0.211	0.430	0.445	0.465	0.385	0.273	0.378	0.410	0.959	1.000			
(18) N65 22.00–23.00 hours	0.124	0.233	0.379	0.400	0.336	0.452	0.283	0.225	0.407	0.410	0.423	0.325	0.139	0.242	0.448	0.708	0.737	1.000		
(19) N60 23.00–24.00 hours	0.096	0.185	0.333	0.335	0.304	0.394	0.374	0.236	0.348	0.349	0.384	0.323	0.130	0.219	0.333	0.372	0.385	0.550	1.000	
(20) N60 24.00–04.30 hours	0.117	0.161	0.238	0.238	0.204	0.281	0.239	0.317	0.250	0.244	0.256	0.239	0.088	0.147	0.264	0.295	0.306	0.318	0.335	1.000

reduction in aircraft noise, or respite, defined as a *scheduled* relief from aircraft noise for a period of time.²² While the minimum noticeable difference in sound level for humans is often regarded as 3 dB, one report suggests that a relief period should provide at least 5–6 dB reduction for people to notice a difference in sound level, and 7–8 dB to provide a significant break from aircraft noise.²³

We considered three ways to identify these reliefs and/or respite periods:

1. Using areas affected by documented flight changes, especially those designed to provide respite.
2. Defining an arbitrary high and low noise level and number of days experienced for each of these:
 - a. For morning periods;
 - b. For afternoon periods;
 - c. Comparing morning and afternoon periods on the same day.
3. Statistically, using CoV of noise for each postcode.

Previous studies on relief/respite

We identified one study by Beghelli *et al.* that examined the effect of respite on medical costs.²⁴ This study used a natural experiment of the Early Morning Arrival Trial, which was implemented from 5 November 2012 to 31 March 2013 to provide noise respite to certain communities living near Heathrow airport. The trial identified four exclusion zones (two to the east and two to the west of Heathrow). This trial designated four exclusion zones (two to the east and two to the west of Heathrow). During the trial, each week, night and early morning (23.30–06.00 hours) aircraft movement was rerouted from one set of air traffic exclusion zones to non-exclusion areas. This reroute alternated weekly between two sets of air traffic exclusion zones. Beghelli *et al.* showed that the trial was associated with a 5.8% decrease in spending on central nervous system treatment, which included the treatment of sleep loss, concentration deficits and other stress-related illnesses in quiet set of zones, compared with control zones, which saw no change in flight movement during the period.²⁴ However, there was a non-statistically significant *increase* in medication for CVD during the trial.²⁴

We did not have information on flight trials to include in our study. However, the trial exclusion zones considered in Beghelli *et al.* were relatively small, with each zone measuring only 1 nautical mile (1.15 miles) in width²⁵ (our study area was 97 × 47 km) and the period was relatively short (5 months – we had 5 years of data). Linking these trial areas is likely to provide an insufficient sample size for health analyses, especially when using binary outcomes indicating more severe disease (hospital admissions and mortality data).

A priori identification of aircraft noise variability based on respite criteria

Using our daily aircraft noise data, we experimented with using arbitrarily defined cut-off noise levels and numbers of days affected to define areas with detectable noise level changes compared with control areas with much more constant levels of noise, with the aim of seeing whether we could identify sufficient postcodes for health analyses.

We attempted to establish criteria for selecting postcodes (relief group) that were exposed to loud aircraft noise on a significant number of days per year but also had a significant number of relatively quiet days. The difference between loud and quiet noise levels was chosen to be 5 dB in order for people to be able to detect the change in sound level according to a previous report on respite and relief periods.²³ We then modified the selection criteria to identify two control groups of postcodes with varying numbers of days exposed to loud aircraft noise. The number of postcodes belonging to each group and their average noise levels were computed and presented in [Tables 7–9](#).

Since our data split a day into eight time bands, we first focused on the morning shoulder period (06.00–07.00 hours), which was typically the noisiest period of the day, as shown in the preceding

section. We arbitrarily identified a relief group that consisted of postcodes with over 100 days of exposure to noise levels above 55 dB and 100 days of exposure to noise levels below 50 dB per year. We chose 55 and 50 dB as our thresholds because a difference of at least 5 dB in sound level is required for people to consciously notice it and also because these are moderate levels of noise at or above mean levels for each time period ([Table 7](#)). The first control group consisted of postcodes that were exposed to aircraft noise during morning shoulder period for at least 100 days below 50 dB but between 100 and 200 days above 55 dB during the same period. The second control group includes postcodes that have at least 200 days that were exposed to above 55 dB during morning shoulder period. [Table 7](#) displays the number of postcodes in each group per year, together with their average decibel levels. Between 23,439 and 25,679 postcodes meet the relief group criteria. However, these numbers represent only about 17% of the total number of postcodes. There were between 2084 and 5065 and 18,374 and 22,235 postcodes that met the criteria for control groups I and II. These numbers nevertheless remain small.

We experimented with various noise thresholds and days to identify postcodes with significant variations in aircraft noise levels. In panel 2 of [Table 7](#), instead of 100 days as the relief group identification criterion, we used 50 days. Similarly, we used 100 days as the cut-off for control group II, as opposed to 200 days. In panel 3 of the same table, we adopted a 7 dB noise difference as opposed to the 5 dB difference used in panel 1. Results indicate that the number of postcodes that meet the criteria in panel 2 and 3 is comparable to that in panel 1.

We used the same method to identify the relief group and the two control groups, but with afternoon noise levels (15.00–19.00 hours). [Table 8](#) shows the results, which demonstrate that similarly small number of postcodes met the criteria.

Finally, we linked morning shoulder (06.00–07.00 hours) noise levels with afternoon (15.00–19.00 hours) noise levels to identify a relief group with a relatively noisy morning shoulder period but a quieter afternoon relative to a control group with significantly noisier morning shoulder and afternoon periods. We identified relief postcodes as those exposed to noise levels above 55 dB during the morning shoulder period but below 50 dB in the afternoon on the same day for at least 100 days per year. The control group consisted of postcodes where morning shoulder period and afternoon noise levels exceeded 55 dB on the same day for at least 200 days per year. [Table 9](#) (top half of panel 1) shows that between 2014 and 2018 there are between 3029 and 5647 postcodes in the relief group, but only between 1655 and 1896 in the control group. We then relaxed the criteria by selecting postcodes that were exposed to noise levels above 55 dB during the morning shoulder period and below 50 dB in the afternoon on the same day but for at least 50 days in a year. The control group comprised all postcodes that were exposed to noise levels above 55 dB during the morning shoulder period and afternoon of the same day for at least 100 days. The results shown in the bottom half of panel 1 of [Table 9](#) display that the number of postcodes in the relief and control groups increased significantly compared with the previous analysis, but the total number remained relatively small. We reapplied the same strategy but increased the noise level difference from 5 to 7 dB to identify relief and control postcodes. The results were presented in panel 2 of [Table 9](#).

In conclusion, using different selection criteria for relief and control postcodes and predefined noise cut-off points to identify areas with large changes in noise levels, we only identified a small proportion of postcodes in the study area (~5000–25,000 postcodes compared with 150,000 overall). Further, some of the predefined control areas had similar or lower noise levels to the areas presumed to have relief periods. We therefore did not conduct health analyses because the small sample size was likely to have been insufficient to detect effects.

Coefficient of variation

The CoV ($SD/mean \times 100$) is a measure of variability that can be used to quantify the day-to-day variability per postcode between 2014 and 2018. A high CoV suggests that daily noise levels within a postcode vary more in a given period.

TABLE 7 Number of postcodes and mean noise levels in predefined relief and two alternative control groups for 06.00–07.00 hours between day changes over 1 year

Difference in dB	Year	Relief area criteria	Postcodes (n)	Mean noise levels (dB)	Control area I criteria	Postcodes (n)	Mean noise levels (dB)	Control area II criteria	Postcodes (n)	Mean noise levels (dB)
Panel 1: 5 dB difference	2014	100+ days > 55 dB AND 100+ < 50 dB days in a year	23,439	51.27	100–200 days > 55 dB AND 100+ days ≤ 50 dB	5065	48.17	200+ days > 55 dB	18,374	52.74
	2015		24,346	52.12		3783	48.94		20,563	53.09
	2016		25,264	51.65		3997	50.32		21,267	52.49
	2017		24,103	54.06		2084	48.74		22,019	54.64
	2018		25,679	51.9		3444	48.17		22,235	52.68
Panel 2: 5 dB difference	2014	50+ days > 55 dB AND 50+ days	25,142	50.81	100–200 days > 55 dB AND 100+ days ≤ 50 dB	5065	48.9	100+ days > 55 dB	23,439	51.27
	2015		26,047	51.72		3783	48.94		24,346	52.12
	2016		26,857	51.21		3997	50.32		25,264	51.65
	2017		26,722	53.35		2084	48.74		24,103	54.06
	2018		27,287	51.46		3444	49.82		25,679	51.9
Panel 2: 7 dB difference	2014	100+ days > 57 dB AND 100+ days	23,439	53.01	100–200 days > 57 dB AND 100+ days ≤ 50 dB	3783	50.39	200+ days > 57 dB	18,374	54.49
	2015		24,346	53.82		3997	50.27		20,563	55.13
	2016		25,264	53.33		2084	52.29		21,267	54.6
	2017		24,103	55.89		3444	50.36		22,019	56.59
	2018		25,679	53.52		5065	48.90		22,235	54.5

TABLE 8 Number of postcodes and their mean noise levels in predefined relief and two alternative control groups for 15.00–19.00 hours between day changes over a year

Difference in dB	Year	Relief area criteria	Postcodes (n)	Mean noise levels (dB)	Control area I criteria	Postcodes (n)	Mean noise levels (dB)	Control area II criteria	Postcodes (n)	Mean noise levels (dB)
Panel 1: 5 dB difference	2014	100+ days > 55 dB AND 100+ days in a year	19,743	52.92	100–200 days > 55 dB AND 100+ days ≤ 50 dB	8424	51.00	200+ days > 55 dB	11,319	57.39
	2015		18,769	53.62		7141	51.65		11,628	57.56
	2016		19,872	53.28		7960	51.34		11,912	57.76
	2017		20,712	52.65		8818	50.71		11,894	57.80
	2018		17,861	53.25		6133	50.89		11,728	57.52
Panel 2: 5 dB difference	2014	50+ days > 55 dB AND 50+ days	23,668	51.42	100–200 days > 55 dB AND 100+ days ≤ 50 dB	8424	51.00	100+ days > 55 dB	19,743	52.92
	2015		23,589	51.48		7141	51.65		18,769	53.62
	2016		23,742	51.51		7960	51.34		19,872	53.28
	2017		24,915	51.19		8818	50.71		20,712	52.65
	2018		22,774	51.41		6133	50.89		17,861	53.25
Panel 2: 7 dB difference	2014	100+ days > 57 dB AND 100+ days	15,265	55.01	100–200 days > 57 dB AND 100+ days ≤ 50 dB	5608	53.29	200+ days > 57 dB	9696	59.68
	2015		14,641	55.57		4833	53.76		9981	59.78
	2016		15,465	55.33		5055	53.44		10,515	59.68
	2017		15,888	54.94		5263	53.1		10,649	59.52
	2018		13,836	55.57		3557	53.34		10,333	59.46

TABLE 9 Number of postcodes and their mean noise levels in pre-defined relief and two alternative control groups, 06.00–07.00 hours compared with 15.00–19.00 hours, linking morning and afternoon period

Difference in dB	Year	Relief area criteria	Postcodes (n)	Period (hours) Mean noise level (dB)	Control area criteria	Postcodes (n)	Period (hours) Mean noise level (dB)
Panel 1: 5 dB difference	2014	100+ days > 55 dB morning 07.00–15.00 hours AND 100+ days ≤ 50 dB afternoon on same day	3029	06.00–07.00: 50.34 15.00–19.00: 48.77	200+ days > 55 dB in morning and afternoon on same day	1655	06.00–07.00: 57.71 15.00–19.00: 58.28
	2015		4009	06.00–07.00: 51.19 15.00–19.00: 48.87		1689	06.00–07.00: 58.55 15.00–19.00: 58.28
	2016		4454	06.00–07.00: 50.60 15.00–19.00: 48.9		1670	06.00–07.00: 58.7 15.00–19.00: 58.51
	2017		5647	06.00–07.00: 50.68 15.00–19.00: 48.7		1896	06.00–07.00: 60.86 15.00–19.00: 58.74
	2018		4090	06.00–07.00: 50.68 15.00–19.00: 47.62		1483	06.00–07.00: 59.18 15.00–19.00: 58.28
	2014	50+ days > 55 dB morning 07.00–15.00 hours AND 50+ days ≤ 50 dB afternoon on same day	6122	06.00–07.00: 50.30 15.00–19.00: 48.96	100+ days > 55 dB morning and afternoon on same day	7375	06.00–07.00: 54.65 15.00–19.00: 53.96
	2015		6884	06.00–07.00: 51.28 15.00–19.00: 49.1		7598	06.00–07.00: 55.39 15.00–19.00: 53.86
	2016		7312	06.00–07.00: 50.69 15.00–19.00: 49.18		7742	06.00–07.00: 54.97 15.00–19.00: 53.91
	2017		12,967	06.00–07.00: 53.15 15.00–19.00: 49.25		7440	06.00–07.00: 57.1 15.00–19.00: 53.64
	2018		8011	06.00–07.00: 51.08 15.00–19.00: 48.59		5811	06.00–07.00: 55.64 15.00–19.00: 54.35

TABLE 9 Number of postcodes and their mean noise levels in pre-defined relief and two alternative control groups, 06.00–07.00 hours compared with 15.00–19.00 hours, linking morning and afternoon period (*continued*)

Difference in dB	Year	Relief area criteria	Postcodes (n)	Period (hours) Mean noise level (dB)	Control area criteria	Postcodes (n)	Period (hours) Mean noise level (dB)
Panel 2: 7 dB difference	2014	100+ days > 57 dB morning 07.00–15.00 hours AND 100+ days ≤ 50 dB afternoon on same day	1235	06.00–07.00: 51.96 15.00–19.00: 50.16	200+ days > 57 dB morning AND afternoon on same day	778	06.00–07.00: 59.08 15.00–19.00: 60.19
	2015		1508	06.00–07.00: 52.89 15.00–19.00: 50.39		769	06.00–07.00: 59.75 15.00–19.00: 60.10
	2016		1748	06.00–07.00: 52.33 15.00–19.00: 50.39		806	06.00–07.00: 60.2 15.00–19.00: 60.28
	2017		2493	06.00–07.00: 54.33 15.00–19.00: 50.11		798	06.00–07.00: 61.54 15.00–19.00: 60.17
	2018		1733	06.00–07.00: 52.07 15.00–19.00: 48.59		658	06.00–07.00: 60.21 15.00–19.00: 59.62
	2014	50+ days > 57 dB morning 07.00–15.00 hours AND 50+ days ≤ 50 dB afternoon on same day	3095	06.00–07.00: 52.31 15.00–19.00: 50.7	100+ days > 57 dB morning AND afternoon on same day	4955	06.00–07.00: 56.56 15.00–19.00: 55.79
	2015		3673	06.00–07.00: 53.01 15.00–19.00: 50.63		4888	06.00–07.00: 57.42 15.00–19.00: 55.78
	2016		4078	06.00–07.00: 52.32 15.00–19.00: 50.55		4905	06.00–07.00: 57.15 15.00–19.00: 55.95
	2017		7481	06.00–07.00: 54.76 15.00–19.00: 50.73		4693	06.00–07.00: 59.14 15.00–19.00: 55.68
	2018		4487	06.00–07.00: 52.83 15.00–19.00: 50.08		3323	06.00–07.00: 58.06 15.00–19.00: 56.87

The 2014–18 descriptive summary of the CoV for daily noise levels by postcode level is presented in [Table 10](#). In this table, for each time band, we calculated the CoV for all seasons (winter, winter transition, summer and summer transition) and additionally by pooled seasonal data across 2014–18 (summer, summer transition, winter and winter transition). Our findings showed that daily noise levels by postcodes varied more during the night. Particularly, 24.00–04.30 hours had the highest mean CoV (67.33–74.16) of all time bands, followed by 04.30–06.00 and 23.00–24.00 hours. The morning shoulder period (06.00–07.00 hours) had the highest mean daily levels, but its mean CoV (15.98–16.83) was the fifth highest among all time bands. Daytime aircraft noise tended to be less variable.

Methods and results: investigating daily aircraft noise and material and health inequality

The hedonic pricing model suggests that aircraft noise is a negative externality that could have a negative impact on housing prices, resulting in noise inequality where the poor are more likely to reside in noisier areas.²⁶ A study relating to London Heathrow airport found no evidence that deprived populations were more likely to be exposed to high aircraft noise levels.²⁷ In fact, the study found that individuals with the highest household income, white ethnicity, and with the lowest area-level income deprivation were more likely to live within a 50 dB contour of aircraft noise. A review examining social inequalities in noise exposure from all sources also found a mixed relationship between deprivation and noise exposure.²⁸ The question of whether aircraft noise may be associated with deprivation is therefore unclear and is likely to vary between airports and countries.

One issue to consider is that deprivation is a potentially multidimensional concept, encompassing numerous facets of an individual's life throughout their lifetime.²⁹ There is limited evidence on the relationship between aircraft noise and non-material deprivation, particularly health inequality, which may be directly linked to health outcomes.

In light of this finding, the purpose of this section of the study was to investigate the relationship between aircraft noise and material and health deprivation.

Deprivation

Given that deprivation has many different aspects, we focused on two: material and health deprivation. We measured material and health deprivation using three variables: Carstairs index of multiple deprivation [census output areas level (COA), 2011 only], fuel poverty rate [lower-layer super output areas level (LSOA), 2014–18] and avoidable death rate per 100,000 [local authority district level (LAD), 2014–18].

- Carstairs index is a commonly used area-level measure of material deprivation in health studies.³⁰ It was calculated using four variables from the 2011 Census, including male unemployment, low social class, non-car ownership and overcrowding. This variable has the highest spatial resolution among the three deprivation indicators chosen for this study, due to its geography being COA (the highest spatial resolution of English Census geography of average population of 310 individuals). This indicator is time invariant as only 2011 values were available. Data were obtained via the UK Data Archive (link: www.data-archive.ac.uk, accessed 23 November 2022).
- Annual fuel poverty rate is used to measure the percentage of households that were unable to maintain standard thermal comfort and safety.³¹ Fuel poverty has been increasingly recognised as a distinct form of social and health inequality.³² It has been hypothesised that cold may be associated with excess winter deaths.³³ A cold home due to fuel poverty has been linked to respiratory problems, arthritis and rheumatism in people of all ages, as well as mental health problems in adolescents.³⁴ This indicator is annual, covering the period 2014–18. The geographic level is LSOA level (Census geography category with average population of 1500 individuals) and covers the period 2014–18. We extracted fuel poverty from UK annual fuel poverty statistics.³⁵

TABLE 10 Descriptive summary of CoV for daily noise levels by postcode level 2014–18

Time band	Season	(1)	(2)	(3)	(4)	(5)
		N	Mean	SD	Min	Max
04.30–06.00 hours	Four seasons	164,012	44.80	16.85	3.301	112.9
	Winter	164,012	44.04	16.91	2.962	109.2
	Winter transition	164,012	46.52	18.07	3.420	126.5
	Summer	164,012	45.92	18.91	3.424	128.0
	Summer transition	164,012	43.91	19.66	3.339	142.0
06.00–07.00 hours	Four seasons	164,012	16.59	7.980	1.655	53.11
	Winter	164,012	16.51	7.952	1.641	52.79
	Winter transition	164,012	16.49	8.004	1.314	53.56
	Summer	164,012	16.83	8.168	1.239	54.25
	Summer transition	164,012	15.98	7.502	1.238	54.28
07.00–15.00 hours	Four seasons	164,012	10.43	4.798	1.194	22.31
	Winter	164,012	10.35	4.731	1.202	22.00
	Winter transition	164,012	10.46	4.901	1.119	22.95
	Summer	164,012	10.59	5.018	1.119	23.47
	Summer transition	164,012	9.728	4.352	1.088	22.71
15.00–19.00 hours	Four seasons	164,012	10.62	4.849	1.315	23.44
	Winter	164,012	10.62	4.825	1.324	23.27
	Winter transition	164,012	10.54	4.913	1.271	23.82
	Summer	164,012	10.60	4.968	1.252	24.15
	Summer transition	164,012	9.892	4.364	1.269	22.54
19.00–22.00 hours	Four seasons	155,951	10.29	4.155	1.388	23.56
	Winter	155,951	10.13	4.074	1.366	23.11
	Winter transition	155,951	10.42	4.349	1.387	24.24
	Summer	155,951	10.61	4.487	1.443	25.08
	Summer transition	155,951	10.05	4.077	1.387	24.22
22.00–23.00 hours	Four seasons	164,012	24.07	37.24	2.126	210.9
	Winter	164,012	24.20	37.23	2.167	210.8
	Winter transition	164,012	23.81	37.26	1.929	210.8
	Summer	164,012	23.62	37.30	1.858	210.7
	Summer transition	164,012	23.47	37.32	1.839	210.6
23.00–24.00 hours	Four seasons	164,012	45.64	35.95	8.104	255.4
	Winter	164,012	48.20	36.03	8.769	263.0
	Winter transition	164,012	41.10	35.86	6.454	241.3
	Summer	164,012	39.12	35.97	5.863	241.2
	Summer transition	164,012	38.08	35.99	5.730	239.3

continued

TABLE 10 Descriptive summary of CoV for daily noise levels by postcode level 2014–18 (*continued*)

Time band	Season	(1)	(2)	(3)	(4)	(5)
		N	Mean	SD	Min	Max
24.00–04.30 hours	Four seasons	164,012	72.13	17.84	20.16	238.4
	Winter	164,012	74.16	18.89	19.26	290.6
	Winter transition	164,012	69.09	16.76	20.46	185.2
	Summer	164,012	68.75	16.79	20.03	169.7
	Summer transition	164,012	67.33	17.02	7.514	299.5

- We used yearly avoidable death rate per 100,000 to measure health inequality. Mortality is an outcome that can be clinically quantified; avoidable mortality is amenable to policy intervention.³⁶ Avoidable death rate could therefore be used to capture the geographical disparity in health.³⁶ The definition of avoidable death rate is available from the Office for National Statistics.³⁷ The data were at LAD level (mean population of approximately 179,361.6 per LAD), covering each year 2014–18. We downloaded the data from the Office for National Statistics.³⁸

Confounders

We adjusted for the quintiles of percentage of non-white population per LAD, considering that ethnic concentration may be related to both deprivation and aircraft noise levels. These data were obtained from Nomis (www.nomisweb.co.uk, accessed 13 February 2023).

Statistical analyses

Since noise exposure levels were calculated daily, serial correlation is a concern. We specified a random effects model with autoregressive first-order autoregression model disturbance to estimate the association between daily noise levels and quintiles of deprivation.

The equations are specified as:

$$noise_{it} = carstairs_j + year_t + ethnic_{it} + u_i + e_{it} \quad (1)$$

where i represents individual postcode, j represents individual output areas, k represents individual LSOA, l represents individual LAD, t represents year; $noise_{it}$, $carstairs_j$ and $ethnic_{it}$ represents daily noise levels (continuous), quintiles of Carstairs index and quantiles of percentage ethnic minority population; $year_t$ is the year fixed effect; u_i is the random heterogeneity and e_{it} is error term.

$$noise_{it} = avoid_{it} + year_t + ethnic_{it} + u_i + e_{it} \quad (2)$$

where i represents individual postcode, j represents individual output areas, k represents individual LSOA, l represents individual LAD and t represents year; $noise_{it}$, $avoid_{it}$ and $ethnic_{it}$ represent daily noise levels (continuous), quintiles of avoidable death rate and quantiles of percentage ethnic minority population; $year_t$ is the year fixed effect; u_i is the random heterogeneity and e_{it} is error term.

$$noise_{it} = fuelpov_{kt} + year_t + ethnic_{it} + u_i + e_{it} \quad (3)$$

where i represents individual postcode, j represents individual output areas, k represents individual LSOA, l represents individual LAD and t represents year; $noise_{it}$, $carstairs_j$, $avoid_{it}$, $fuelpov_{kt}$ and $ethnic_{it}$ represent daily noise levels (continuous), quintiles of fuel poverty rate and quantiles of percentage ethnic minority population; $year_t$ is the year fixed effect; u_i is the random heterogeneity and e_{it} is error term.

$$noise_{it} = carstairs_j + avoid_{it} + fuelpov_{kt} + year_t + ethnic_{it} + u_i + e_{it} \quad (4)$$

where i represents individual postcode, j represents individual output areas, k represents individual LSOA, l represents individual LAD and t represents year; $noise_{it}$, $fuelpov_{kt}$ and $ethnic_{lt}$ represent daily noise levels (continuous), quintiles of Carstairs index, quintiles of avoidable death rate, quintiles of fuel poverty rate and quintiles of percentage ethnic minority population; $year_t$ is the year fixed effect; u_i is the random heterogeneity and e_{it} is error term.

Each postcode uniquely belongs to an output area, a lower LSOA and a LAD, which enables us to link data.

Our dependent variables included the four noise metrics: LAeq24, Lnight, Leve and Lday.

We conducted four regressions per noise metric. The first regression of each metric included quintiles of Carstairs index and percentage non-white ethnicity. Models 2 and 3 replaced Carstairs index with avoidable death rate and fuel poverty rate, respectively. Model 4 included quintiles for all three measures of deprivation and percentage non-white ethnicity.

All analyses were conducted in Stata using module xtregar.³⁹

Results

Table 11 shows the descriptive summary of measures of deprivation and percentage non-white ethnicity in the analysis.

Table 12 illustrates the pairwise correlations between variables involved in analysis. The noise correlation coefficients between LAeq24, Lday, Leve and Lnight ranged between $r = 0.68$ and 0.98 . The correlation between Lnight and Leve ($r = 0.68$) was the lowest among all pairs, whereas the correlation between Lday and LAeq24 ($r = 0.98$) was the highest. For deprivation variables, we used raw values rather than quintiles. The correlation between Carstairs index and both of avoidable death rate and fuel poverty was moderate ($r \sim 0.4$), while that between fuel poverty rate and avoidable death rate was particularly weak ($r = 0.08$). There was a fairly weak relationship between each pair of the three deprivation variables and area percentage non-white ethnicity in our data ($r 0.08$ – 0.49), with the correlation between the Carstairs index and the percentage non-white ethnicity being the strongest ($r = 0.49$).

Tables 13 and **14** demonstrate the main results from regressions. The dependent variables were Lday (07.00–19.00 hours) and Leve (19.00–23.00 hours) in **Table 13**, and Lnight (23.00–07.00 hours) and LAeq24 (24-hour average) in **Table 14**. In models 1–3, we separately regressed the association between one measure of deprivation and aircraft noise levels while adjusting for quintiles of percentage non-white ethnicity. The results of these models consistently demonstrated that almost all quintiles of the Carstairs index, avoidable death rate and fuel poverty rate (except for Q5 of the fuel poverty rate) had significant and positive coefficients, regardless of the noise metrics being examined. This evidenced that postcodes near Heathrow airport with the least material or health deprivation experienced the lowest daily noise levels between 2014 and 2018.

TABLE 11 Descriptive summary of measures of deprivation and percentage non-white ethnicity

Variables	(1)	(2)	(3)	(4)	(5)
	N	Mean	SD	Min	Max
Avoidable death rate per 10,000 persons	284,476,323	133.26	25.1	78	209.9
Carstairs index	284,476,323	0.94	3.1	–4.88	28.31
Fuel poverty (%)	284,476,323	10.18	3.63	1.8	29.6
Non-white (%)	284,476,323	36.26	14.51	4.4	68.9

TABLE 12 Pairwise correlations between noise metrics, deprivation measures and control variable

Variables	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
(1) Lday	1							
(2) Leve	0.82	1						
(3) LAeq24	0.98	0.9	1					
(4) Lnight	0.78	0.68	0.82	1				
(5) Carstairs index	-0.02	-0.03	-0.02	0.04	1			
(6) Avoidable death rate	0.15	0.15	0.19	0.23	0.42	1		
(7) Fuel poverty rate	-0.07	-0.09	-0.08	-0.03	0.36	0.08	1	
(8) Non-white (%)	-0.13	-0.14	-0.15	-0.07	0.49	0.41	0.36	1

However, which quintile of deprivation (Q1 least deprived, Q5 least deprived) was the noisiest depended on the deprivation measures and noise metrics used. During the day (07.00–19.00 hours), Q2 of Carstairs index, Q4 of avoidable death rate and Q4 of fuel poverty rate had the highest levels of noise. In the evening (19.00–23.00 hours), Q2 of the Carstairs index, Q4 of the avoidable death rate and Q2 of fuel poverty rate had the highest levels of noise pollution. It is interesting to note, the quintile with the highest levels of aircraft noise exposure at night (23.00–07.00 hours) was Q5, for both the Carstairs index and the avoidable death rate, as well as Q3 for the fuel poverty rate.

There were two interesting exposure–response patterns in relation to this. First, among the three indicators of deprivation, avoidable death rate had the most noticeable gradient. Column (2) of [Table 13](#) demonstrates Q2 and Q3 postcodes were exposed to slightly higher noise levels, whereas Q4 and Q5 postcodes were exposed to significantly higher noise levels during daytime (coefficients: Q2 – 0.08, Q3 – 0.48, Q4 – 0.76 and Q5 – 0.69). The second is that the exposure–response relationship between night-time aircraft noise levels and deprivation was more pronounced than during the day and evening. This gradient was particularly clear when we paired night-time noise with avoidable death rate (coefficients: Q2 – 0.41, Q3 – 1.22, Q4 – 1.59 and Q5 – 1.66). While the relationship did not appear to be linear, it supported an observation that postcodes in local authorities with higher avoidable death rates were more likely to be exposed to a higher level of aircraft noise at night.

We found a stronger association between deprivation and daily aircraft noise at night than during the day or evening in postcodes near Heathrow airport. In comparison with postcodes in Q1, those in Q2–Q5 of the Carstairs index were exposed to noise levels that were 2.66, 3.13, 2.79 and 3.18 dB higher at night, but only 1.02, 0.58, 0.09 and 0.40 dB higher during daytime, and 1.02, 0.68, 0.22 and 0.48 dB higher during evening. The same conclusion is supported by the results that LAeq24 (mean sound levels over the 24 hours) had very similar number of observations as Lnight (*N* observations: Lnight – 85,847,742 vs. LAeq24 – 82,265,795) but its relationship with deprivation was significantly smaller in size than that of Lnight. The evidence suggests that night-time aircraft noise exposure inequality is of particular concern in postcodes near Heathrow.

Lastly, we found that the fuel poverty rate had a weaker relationship with daily aircraft noise than the Carstairs index and avoidable death rate. There was a negative relationship between the fifth and fourth quintiles of fuel poverty and aircraft noise during some periods.

TABLE 13 The association between deprivation measures and 24-hour and daytime aircraft noise

Dependent variables	Lday				Leve			
	Model 1	Model 2	Model 3	Model 4	Model 1	Model 2	Model 3	Model 4
<i>Carstairs index Q1 – least deprived (base)</i>								
Q2	1.02*** (0.05)			0.90*** (0.05)	1.02*** (0.05)			0.90*** (0.04)
Q3	0.58*** (0.04)			0.35*** (0.04)	0.68*** (0.04)			0.45*** (0.04)
Q4	0.09** (0.04)			-0.21*** (0.04)	0.22*** (0.04)			-0.08** (0.04)
Q5	0.40*** (0.04)			-0.01 (0.04)	0.48*** (0.04)			0.11*** (0.04)
<i>Avoidable death rate Q1 – least deprived (base)</i>								
Q2		0.08*** (0.00)		0.08*** (0.00)		0.12*** (0.00)		0.13*** (0.00)
Q3		0.48*** (0.01)		0.49*** (0.01)		0.38*** (0.01)		0.38*** (0.01)
Q4		0.76*** (0.01)		0.76*** (0.01)		0.71*** (0.01)		0.71*** (0.01)
Q5		0.69*** (0.01)		0.70*** (0.01)		0.69*** (0.01)		0.69*** (0.01)
<i>Fuel poverty rate Q1 – least deprived (base)</i>								
Q2			0.02*** (0.00)	0.02*** (0.00)			0.15*** (0.00)	0.15*** (0.00)
Q3			0.03*** (0.00)	0.03*** (0.00)			0.12*** (0.00)	0.12*** (0.00)
Q4			0.04*** (0.00)	0.04*** (0.00)			0.12*** (0.00)	0.12*** (0.00)
Q5			-0.02*** (0.00)	-0.01*** (0.00)			0.00 (0.00)	0.01*** (0.00)
<i>Non-white ethnicity (%) Q1 (base)</i>								
Q2	0.15*** (0.01)	0.17*** (0.01)	0.16*** (0.01)	0.18*** (0.01)	-0.48*** (0.00)	-0.45*** (0.00)	-0.46*** (0.00)	-0.43*** (0.00)
Q3	-0.12*** (0.01)	-0.14*** (0.01)	-0.12*** (0.01)	-0.14*** (0.01)	-0.63*** (0.01)	-0.65*** (0.01)	-0.62*** (0.01)	-0.64*** (0.01)
Q4	-0.57*** (0.01)	-0.56*** (0.01)	-0.57*** (0.01)	-0.55*** (0.01)	-1.13*** (0.01)	-1.11*** (0.01)	-1.12*** (0.01)	-1.10*** (0.01)

continued

TABLE 13 The association between deprivation measures and 24-hour and daytime aircraft noise (*continued*)

Dependent variables	Lday				Leve			
	Model 1	Model 2	Model 3	Model 4	Model 1	Model 2	Model 3	Model 4
Q5	-0.78*** (0.01)	-0.85*** (0.01)	-0.78*** (0.01)	-0.84*** (0.01)	-1.49*** (0.01)	-1.57*** (0.01)	-1.47*** (0.01)	-1.55*** (0.01)
Constant	43.93*** (0.03)	43.92*** (0.01)	44.29*** (0.01)	43.78*** (0.03)	42.78*** (0.03)	42.87*** (0.01)	43.14*** (0.01)	42.57*** (0.03)
Observations (n)	280,458,080	280,458,080	280,458,080	280,458,080	268,178,754	268,178,754	268,178,754	268,178,754
Postcodes (n)	162,004	162,004	162,004	162,004	154,173	154,173	154,173	154,173
Autocorrelation	AR1	AR1	AR1	AR1	AR1	AR1	AR1	AR1

*** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

Note

Standard errors in parentheses.

Models 1–3 only use one measure of deprivation: quintiles of Carstairs index in model 1, quintiles of avoidable death rate in model 2 and quintiles of fuel poverty in model 3. Model 4 includes all three measures of deprivation in the regression. All models have controlled for quintiles of percentage non-white ethnicity.

TABLE 14 The association between deprivation measures and evening and night-time aircraft noise

Dependent variables	Lnight				LAeq24			
	Model 1	Model 2	Model 3	Model 4	Model 1	Model 2	Model 3	Model 4
<i>Carstairs index Q1 – least deprived (base)</i>								
Q2	2.66*** (0.06)			2.42*** (0.05)	1.12*** (0.05)			0.95*** (0.05)
Q3	3.13*** (0.05)			2.68*** (0.05)	0.99*** (0.05)			0.66*** (0.04)
Q4	2.79*** (0.05)			2.23*** (0.04)	0.57*** (0.05)			0.15*** (0.04)
Q5	3.18*** (0.05)			2.45*** (0.04)	0.81*** (0.04)			0.26*** (0.04)
<i>Avoidable death rate Q1 – least deprived (base)</i>								
Q2		0.41*** (0.01)		0.41*** (0.01)		0.17*** (0.01)		0.18*** (0.01)
Q3		1.22*** (0.01)		1.13*** (0.01)		0.62*** (0.01)		0.63*** (0.01)
Q4		1.59*** (0.01)		1.47*** (0.01)		1.20*** (0.01)		1.20*** (0.01)
Q5		1.66*** (0.02)		1.54*** (0.02)		1.11*** (0.02)		1.12*** (0.02)

TABLE 14 The association between deprivation measures and evening and night-time aircraft noise (continued)

Dependent variables	Lnight				LAeq24			
	Model 1	Model 2	Model 3	Model 4	Model 1	Model 2	Model 3	Model 4
Fuel poverty rate Q1 – least deprived (base)								
Q2			0.16*** (0.01)	0.14*** (0.01)			0.07*** (0.01)	0.06*** (0.01)
Q3			0.23*** (0.01)	0.22*** (0.01)			0.08*** (0.01)	0.07*** (0.01)
Q4			0.13*** (0.01)	0.10*** (0.01)			0.08*** (0.01)	0.07*** (0.01)
Q5			-0.08*** (0.01)	-0.11*** (0.01)			-0.04*** (0.01)	-0.03*** (0.01)
Non-white ethnicity (%) Q1 (base)								
Q2	-0.51*** (0.01)	-0.43*** (0.01)	-0.35*** (0.01)	-0.52*** (0.01)	-0.37*** (0.01)	-0.38*** (0.01)	-0.34*** (0.01)	-0.37*** (0.01)
Q3	-0.42*** (0.01)	-0.40*** (0.01)	-0.25*** (0.01)	-0.53*** (0.01)	-0.53*** (0.01)	-0.62*** (0.01)	-0.50*** (0.01)	-0.61*** (0.01)
Q4	-0.95*** (0.01)	-0.89*** (0.01)	-0.78*** (0.01)	-1.04*** (0.01)	-1.02*** (0.01)	-1.05*** (0.01)	-0.99*** (0.01)	-1.04*** (0.01)
Q5	-1.10*** (0.02)	-1.15*** (0.02)	-0.88*** (0.02)	-1.28*** (0.02)	-1.24*** (0.02)	-1.43*** (0.02)	-1.18*** (0.02)	-1.42*** (0.02)
Constant	33.33*** (0.04)	34.94*** (0.02)	35.69*** (0.02)	32.91*** (0.04)	42.31*** (0.04)	42.48*** (0.02)	42.95*** (0.02)	42.07*** (0.03)
Observations (n)	85,847,742	85,847,742	85,847,742	85,847,742	82,265,795	82,265,795	82,265,795	82,265,795
Postcodes (n)	162,003	162,003	162,003	162,003	154,173	154,173	154,173	154,173
Autocorrelation	AR1	AR1	AR1	AR1	AR1	AR1	AR1	AR1

*** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

Notes

Standard errors in parentheses.

Models 1–3 only use one measure of deprivation: quintiles of Carstairs index in model 1, quintiles of avoidable death rate in model 2 and quintiles of fuel poverty in model 3. Model 4 includes all three measures of deprivation in the regression. All models have controlled for quintiles of percentage non-white ethnicity.

The above interpretation of the results was based on models that included only one measure of deprivation, but our conclusions held when all deprivation measures were included (model 4).

Methods and results: short-term impact of aircraft noise on cardiovascular morbidity and mortality

One of the main aims of our study was to evaluate short-term impact of aircraft noise on cardiovascular morbidity and mortality. We used a time-stratified case-crossover study design, in which the days when an event of interest occurred are matched with control days within the same month and on the

same day of the week.^{40,41} This individual-level design naturally adjusts for all time-invariant or slowly time-varying confounders, including sex, smoking behaviour and genetic factors. It uses all cases in the population without the need to recruit additional controls. The case-crossover design is useful in assessing the acute impact of a transient risk factor with minimal bias and has been used widely in environmental epidemiology, predominantly in temperature and air pollution studies as well as aircraft noise.^{42,43}

Health outcomes data

All hospital episodes and deaths due to primary CVD in the study area from 1 January 2014 to 31 December 2018 were included. We extracted postcode data on all hospital episodes and deaths from the Hospital Episode Statistics from NHS Digital and the mortality data from the Office for National Statistics held by the UK Small Area Health Statistics Unit at Imperial College London. Data were obtained for all events with primary cause of admission or death due to stroke [International Classification of Diseases version 10 (ICD-10) codes I61, I63–I64], coronary heart disease (CHD; ICD-10 I20–I25) and CVD (ICD-10 Chapter I) and linked to postcode-level noise estimates. Time of hospital episode and death were not available. The study was covered by national research ethics approval from the London – South East Research Ethics Committee (reference 17/LO/0846; date of opinion 29 June 2017). Data access to confidential patient information without consent was covered by the Health Research Authority Confidentiality Advisory Group under Regulation 5 of the Health Service (Control of Patient Information) Regulations 2002 (section 251); reference: 20/CAG/0028 (outcome date 24 March 2020, section 251 Register Index Sheet application number A02476).

Confounder data

The environmental confounders included in the models were mean temperature and particulate matter less than 2.5 μm in diameter ($\text{PM}_{2.5}$) concentration. Hourly dry air temperature measurements were captured at three National Oceanic and Atmospheric Administration Integrated Surface Database weather stations within 25 km of the study area. Hourly background measurements of fine $\text{PM}_{2.5}$ were captured by the six UK Automatic Urban and Rural Network sites within 25 km of the study area. Dry air temperature and background $\text{PM}_{2.5}$ concentrations were estimated at each residential postcode using a spatial interpolation technique known as inverse distance-squared weighting.

Individual-level ethnicity data were available for all hospital admissions in the Hospital Episode Statistics data and COA-level Carstairs Index quintile from the 2011 Census was linked to admissions and deaths data. The Carstairs index is a commonly used indicator of material deprivation in health studies.^{30,44} All estimates were also adjusted for the effect of holidays.

Statistical analyses

Patients with multiple cardiovascular episodes (records) per day ($n = 3,018$; 0.07% of cases) had one record on the day randomly selected for inclusion. Each episode record represents a patient being seen by a new clinician, so these may relate to the same spell in hospital. Control periods were matched to case periods within the same year and month on the same day of the week, excluding control days on which an additional cardiovascular episode occurred ($n = 15,856$ controls); 528 cases with no suitable control days were also excluded from analyses. A flowchart of the exclusion criteria and how they affected the numbers of cases and controls is presented in [Figure 3](#).

Conditional logistic regression was used to estimate the OR and 95% CI per 5-dB increase for the metrics L_{day} , L_{eve} , L_{night} , L_{den} and L_{Aeq24} , as well as for eight distinct periods throughout the 24-hour period relating to aircraft flows. We considered all CVD, CHD only and stroke only for both hospital episode and deaths. Estimates were adjusted for mean temperature, $\text{PM}_{2.5}$ concentration and the effect of holidays, as these are variables that change rapidly in time, while long-term confounders were accounted for by the case-crossover study design. Analyses were also stratified by age, sex, ethnicity, deprivation and season to assess effect modification. All analyses were run in R statistical software⁴⁵ using the *Epi* package.⁴⁶

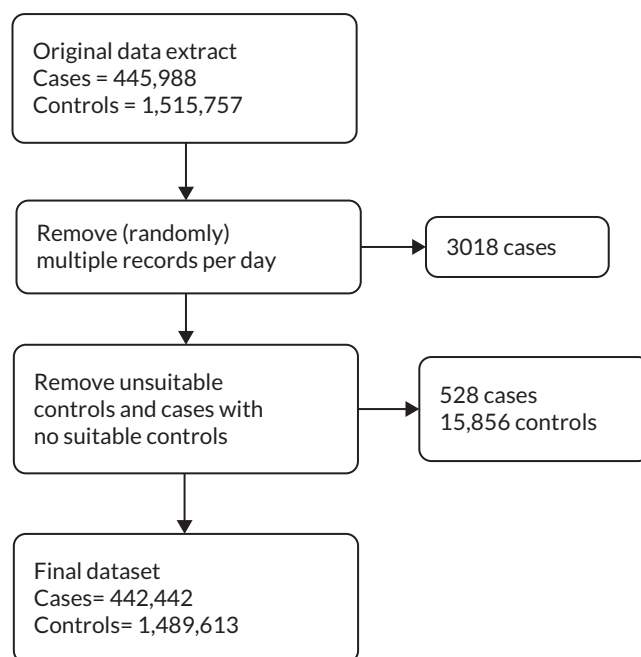


FIGURE 3 Hospital episodes exclusion criteria.

Descriptive results

The descriptive summary of our sample is presented in [Table 15](#) (note that the noise estimates relate to postcodes of cases only. This is a subset of all postcodes and descriptive statistics of noise levels differ from estimates in [Table 5](#) relating to all postcodes; in fact, averages are lower than for all postcodes, likely reflecting spatial variability in characteristics of populations by postcode, such as age of individuals who become cases).

Results for short-term aircraft noise associations with hospital admissions

There was evidence of a small increase in risk for a 5-dB increment in noise during the evening (Leve OR = 1.005, 95% CI 1.000 to 1.010), particularly from 22.00 to 23.00 hours (OR = 1.006, 95% CI 1.002 to 1.010) for all CVD admissions ([Table 16](#)). A similar but not statistically significant pattern was estimated for admissions due to CHD. There was no evidence of an increased risk for hospitalisations due to stroke ([Figure 4](#)).

Stratified analyses for cardiovascular admissions can be seen in [Figure 5](#). After stratifying by age and sex, the effect of aircraft noise on cardiovascular admissions was statistically significant in men over the age of 65 years during the daytime (Lday OR = 1.009, 95% CI 1.001 to 1.019) evening (Leve OR = 1.014, 95% CI 1.005 to 1.019) and, to a lesser extent, in women under the age of 65 years during the evening (Leve OR = 1.012, 95% CI 1.000 to 1.023; [Figure 5](#)). Similarly, the same figure shows, after stratifying by ethnicity, that an effect for hospitalisations due to all CVD was seen in patients who identified as South Asian during the evening hours 22.00–23.00 hours (OR = 1.004, 95% CI 1.002 to 1.028) and as other ethnicity (not South Asian or black) during the evening (Leve OR = 1.007, 95% CI 1.001 to 1.013). There was no significant effect modification by age, sex or ethnicity evident for CHD or stroke. There was also an increase in risk during late night hours among individuals residing in the third and fourth quintiles of deprivation ([Figure 5](#)).

There was also evidence of effect modification by season for all CVD and for CHD. The effect of night-time aircraft noise on all CVD was strongest in the summer (22.00–23.00 hours OR = 1.009, 95% CI 1.001 to 1.017; 23.00–24.00 hours OR = 1.009, 95% CI 1.002 to 1.016) and winter months (22.00–23.00 hours OR = 1.008, 95% CI 1.002 to 1.014; 23.00–24.00 hours OR = 1.009, 95% CI 1.002 to 1.015), and the effect of late afternoon and early evening aircraft noise was only evident in the winter (15.00–19.00 hours

TABLE 15 Descriptive statistics for hospital admissions and deaths due to CVD in the study area, and noise estimates for all CVD event (case) and control days

	Hospital episodes 2014-18 (n = 442,442)					Deaths 2014-18 (n = 49,443)						
	All CVD		CHD		Stroke	All CVD		CHD		Stroke		
	(n)	(%)	(n)	(%)	(n)	(%)	(n)	(%)	(n)	(%)	(n)	(%)
Sex ^a												
Male	256,674	58.0	81,278	69.9	21,367	52.7	26,011	52.6	12,984	61.9	4014	45.7
Female	185,749	42.0	34,941	30.1	19,199	47.3	23,432	47.4	7984	38.1	4771	54.3
Age (years) ^a												
< 65	190,732	43.1	49,936	43.0	11,420	28.2	7267	14.7	3640	17.4	834	9.5
65+	250,705	56.7	66,260	57.0	29,106	71.7	42,176	85.3	17,328	82.6	7951	90.5
Deprivation												
1 (least)	57,060	12.9	15,061	13.0	5080	12.5	7239	14.6	2843	13.6	1349	15.4
2	55,076	12.4	14,307	12.3	5104	12.6	6561	13.3	2630	12.5	1275	14.5
3	72,775	16.4	18,735	16.1	6666	16.4	8582	17.4	3541	16.9	1529	17.4
4	106,033	24.0	27,356	23.5	9454	23.3	11,606	23.5	4984	23.8	2012	22.9
5	151,498	34.2	40,763	35.1	14,263	35.2	15,455	31.3	6970	33.2	2620	29.8
Season												
Summer	110,255	24.9	29,215	25.1	10,095	24.9	11,260	22.8	4728	22.5	2060	23.4
Summer transition	73,835	16.7	19,730	17.0	6731	16.6	7702	15.6	3231	15.4	1333	15.2
Winter	184,625	41.7	47,807	41.1	16,838	41.5	22,365	45.2	9589	45.7	3933	44.8
Winter transition	73,727	16.7	19,470	16.8	6903	17.0	8116	16.4	3420	16.3	1459	16.6
Ethnicity ^b												
South Asian	42,994	9.7	18,049	15.5	2711	6.7						
Black	35,245	8.0	5704	4.9	4197	10.3						

TABLE 15 Descriptive statistics for hospital admissions and deaths due to CVD in the study area, and noise estimates for all CVD event (case) and control days (*continued*)

	Hospital episodes 2014–18 (n = 442,442)					Deaths 2014–18 (n = 49,443)							
	All CVD		CHD		Stroke	All CVD		CHD		Stroke			
	(n)	(%)	(n)	(%)	(n)	(%)	(n)	(%)	(n)	(%)	(n)	(%)	
Other ^a	297,390	67.2	73,658	63.4	27,582	68.0							
Missing	66,813	15.1	18,811	16.2	6077	15.0							
	Case				Control				Case		Control		
Noise estimates (dB)	Mean	SD	Mean	SD	Mean	SD	Mean	SD	Mean	SD	Mean	SD	
24.00–04.30	2.01	5.99	2.04	6.03	2.07	6.08	2.06	6.06					
04.30–06.00	25.80	12.53	25.81	12.52	25.72	12.38	25.71	12.41					
06.00–07.00	40.77	8.80	40.79	8.79	40.59	8.62	40.60	8.62					
07.00–15.00	42.43	6.97	42.42	6.96	42.43	6.83	42.43	6.85					
15.00–19.00	41.93	6.94	41.92	6.94	41.94	6.82	41.92	6.84					
19.00–22.00	41.85	6.87	41.84	6.87	41.79	6.77	41.77	6.78					
22.00–23.00	39.51	7.41	39.49	7.41	39.48	7.41	39.49	7.40					
23.00–24.00	27.92	10.89	27.94	10.90	27.59	10.91	27.64	10.93					

a 19 hospital episodes missing sex, 1005 missing age.

b Ethnicity information not available for mortality data.

c Includes all other non-black and non-South Asian ethnicities, including white and mixed ethnicities.

OR = 1.008, 95% CI 1.001 to 1.014; 19.00–22.00 hours OR = 1.012, 95% CI 1.004 to 1.020). The effect of early morning aircraft noise was strongest in the summer transition months (04.30–06.00 hours OR = 1.018, 95% CI 1.009 to 1.028; 06.00–07.00 hours OR = 1.010, 95% CI 1.003 to 1.017; [Figure 6](#)).

In a sensitivity analysis including only the first hospital episode for the 60.8% of patients with more than one episode ([Figure 7](#)), similar patterns were seen as for the main analyses (see [Figure 4](#) and [Table 16](#)).

Associations of aircraft noise with mortality

There was no evidence of an association between short-term aircraft noise and deaths due to CVD, CHD or stroke, with wide confidence intervals ([Figure 8](#)).

Variability in risk estimates between areas with consistent patterns of noise exposure compared with those with changing patterns of noise exposure

As detailed previously, we used CoV to measure the variability of daily noise levels per postcode between 2014 and 2018. We reran regression analyses stratifying our samples by low (\leq mean) compared with high ($>$ mean) CoV (daily noise levels between 2014 and 2018), to assess the difference in risks of hospitalisation for CVD across regions with low and high availability in noise. The mean CoV for each noise metrics can be found in [Table 10](#). As seen in [Table 17](#), there were increased risks of hospitalisation for CVD throughout the evening hours (19.00–22.00, 22.00–23.00 and 23.00–24.00 hours) in the low CoV group. However, there was a non-significant risk at any time of day in high CoV group.

To explore whether low CoV areas were those with higher noise levels (potentially suggesting high noise and less relief periods), we examined mean noise levels ([Table 18](#)). For the latter two periods, mean noise levels were higher in the low CoV postcodes (41 vs. 37 dB for 22.00–23.00 hours; 31 vs. 24 dB for 23.00–24.00 hours). However, for the period 19.00–22.00 hours, the mean noise levels were 41 dB in low CoV areas compared with 43 dB in high CoV areas. We therefore could not readily infer that a lack of relief periods (or at least some periods of lower noise exposure) was related to the association with hospitalisation.

TABLE 16 Odds ratio and 95% CIs for hospitalisations and deaths due to all CVD per 5-dB increase LAeq. Estimates adjusted for PM_{2.5} concentration, mean temperature and holiday effect

	All CVD	CHD	Stroke
Hosp. episodes			
LAeq24	1.003 (0.998 to 1.008)	1.002 (0.992 to 1.012)	1.003 (0.986 to 1.020)
Lday	1.002 (0.997 to 1.007)	1.002 (0.993 to 1.011)	1.001 (0.986 to 1.017)
Leve	1.005 (1.000 to 1.010)	1.005 (0.996 to 1.014)	0.999 (0.984 to 1.015)
Lnight	0.999 (0.995 to 1.003)	0.997 (0.989 to 1.004)	1.001 (0.989 to 1.014)
Lden	1.001 (0.997 to 1.006)	0.999 (0.989 to 1.009)	1.002 (0.986 to 1.017)
Deaths			
LAeq24	1.002 (0.987 to 1.017)	0.993 (0.970 to 1.017)	0.980 (0.945 to 1.016)
Lday	1.000 (0.986 to 1.014)	0.994 (0.972 to 1.016)	0.984 (0.952 to 1.018)
Leve	1.003 (0.988 to 1.017)	0.990 (0.968 to 1.012)	0.983 (0.950 to 1.017)
Lnight	0.999 (0.988 to 1.011)	1.007 (0.989 to 1.025)	0.981 (0.955 to 1.008)
Lden	1.001 (0.987 to 1.016)	0.997 (0.975 to 1.020)	0.969 (0.936 to 1.003)

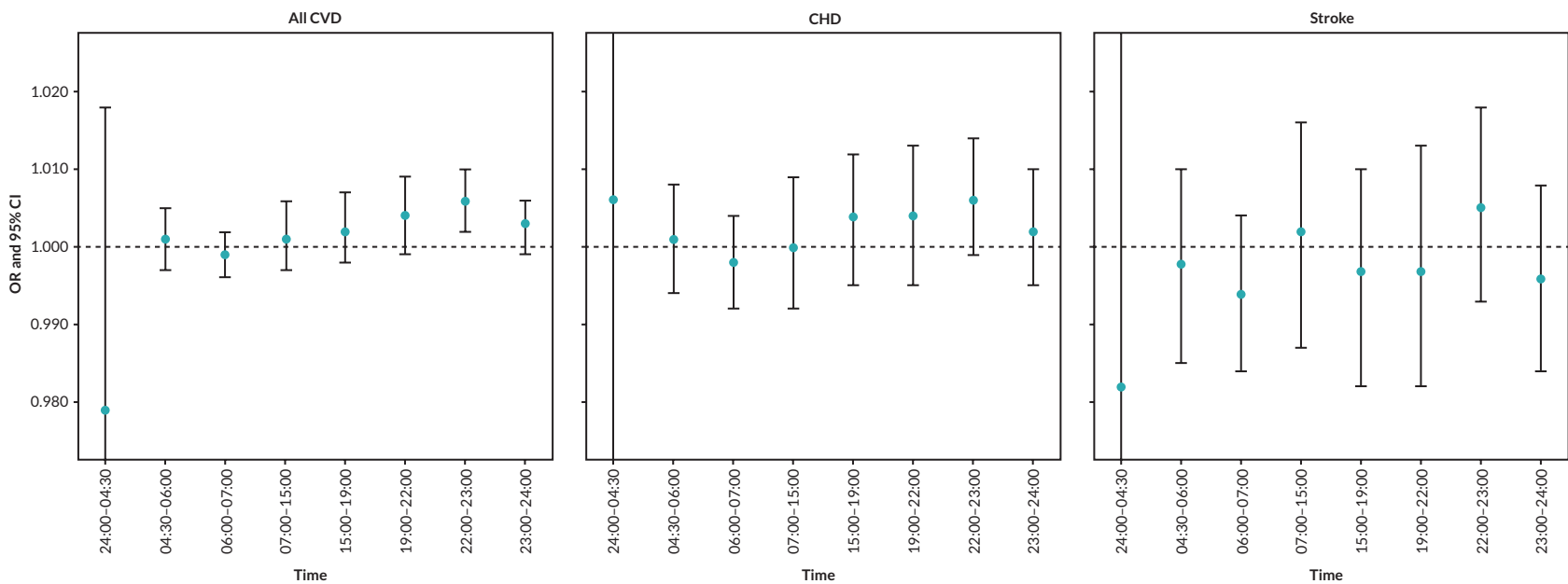


FIGURE 4 Odds ratios and 95% CIs for hospitalisations due to all CVD, CHD and stroke per 5-dB increase LAeq at defined time points throughout the day, evening and night. Estimates adjusted for PM_{2.5} concentration, mean temperature and holiday effect.

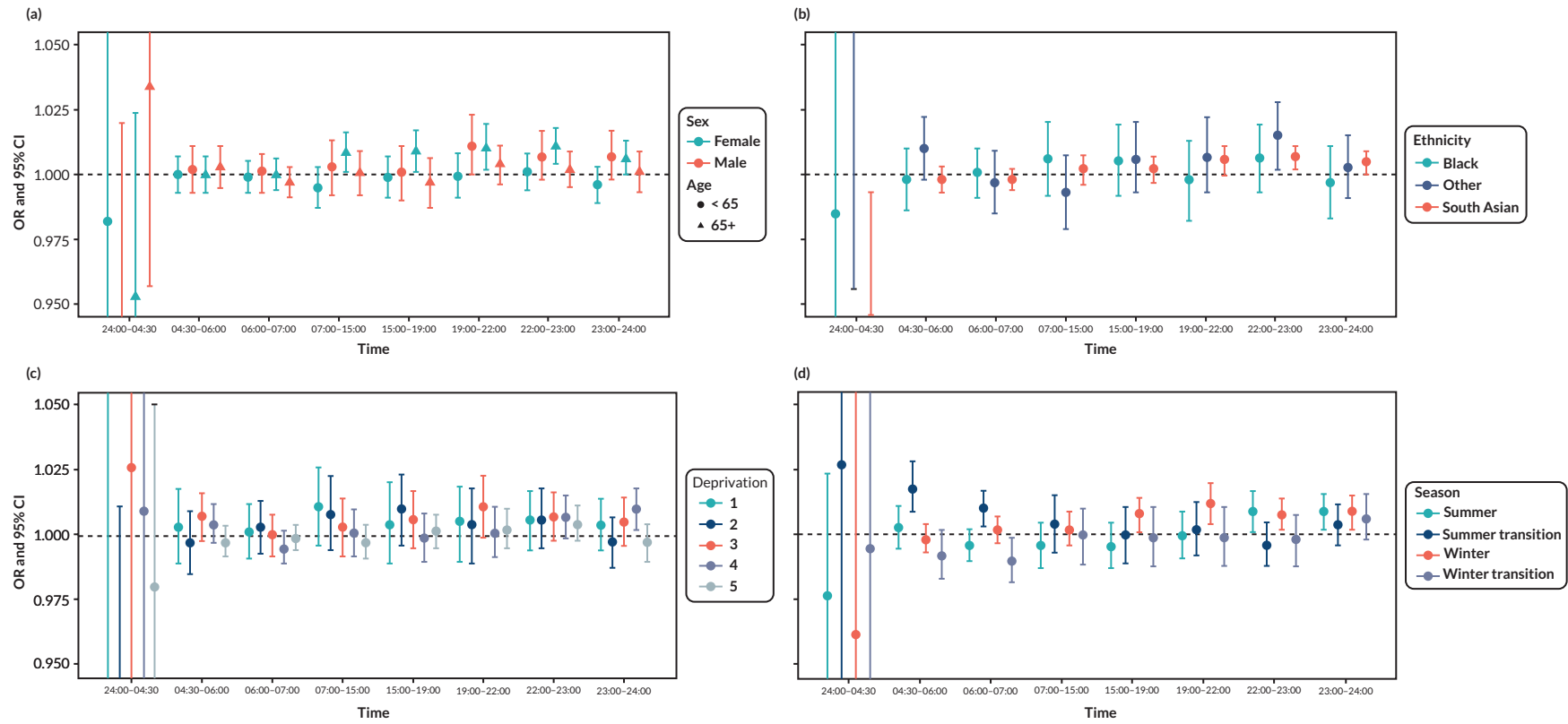


FIGURE 5 Odds ratios and 95% CIs for hospitalisations due to all CVD per 5-dB increase LAeq, stratified by (a) age-sex, (b) ethnicity, (c) deprivation and (d) season. Estimates adjusted for PM_{2.5} concentration, mean temperature and holiday effect.

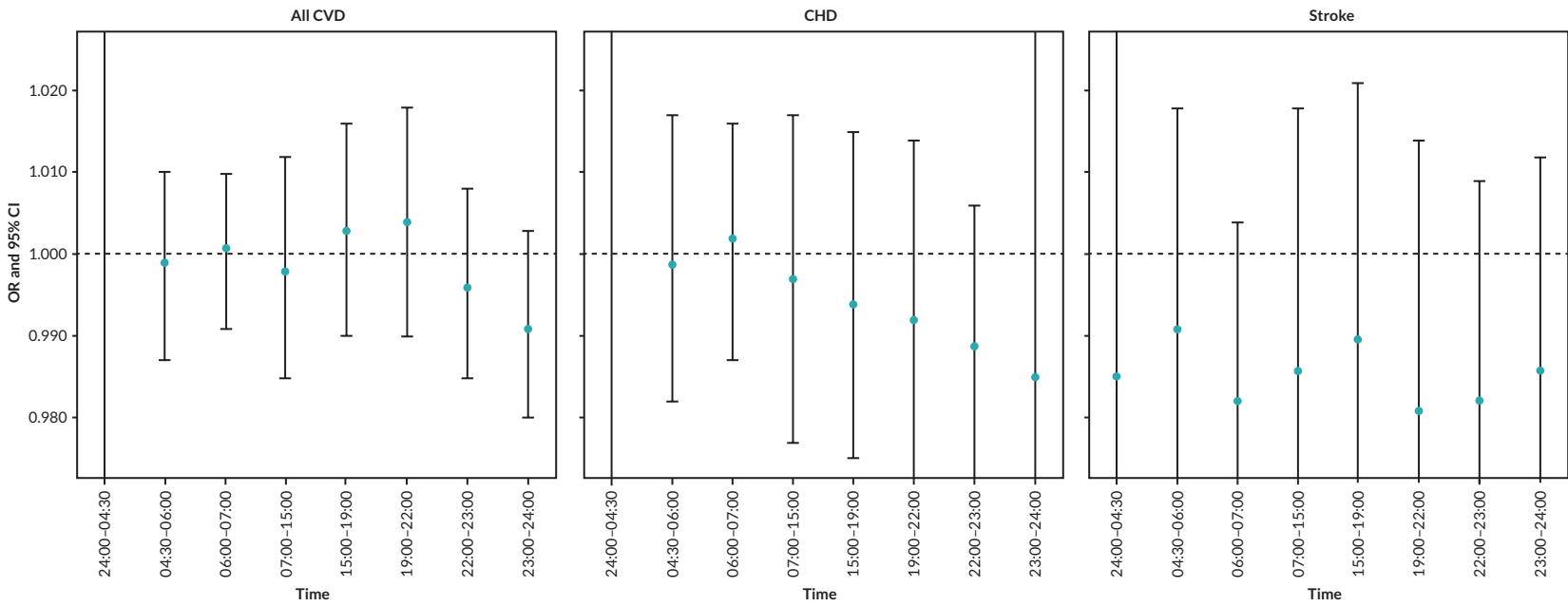


FIGURE 6 Odds ratio and 95% CIs for hospitalisations CHD (right) per 5-dB increase LAeq, stratified by season. Estimates adjusted for PM_{2.5} concentration, mean temperature and holiday effect.

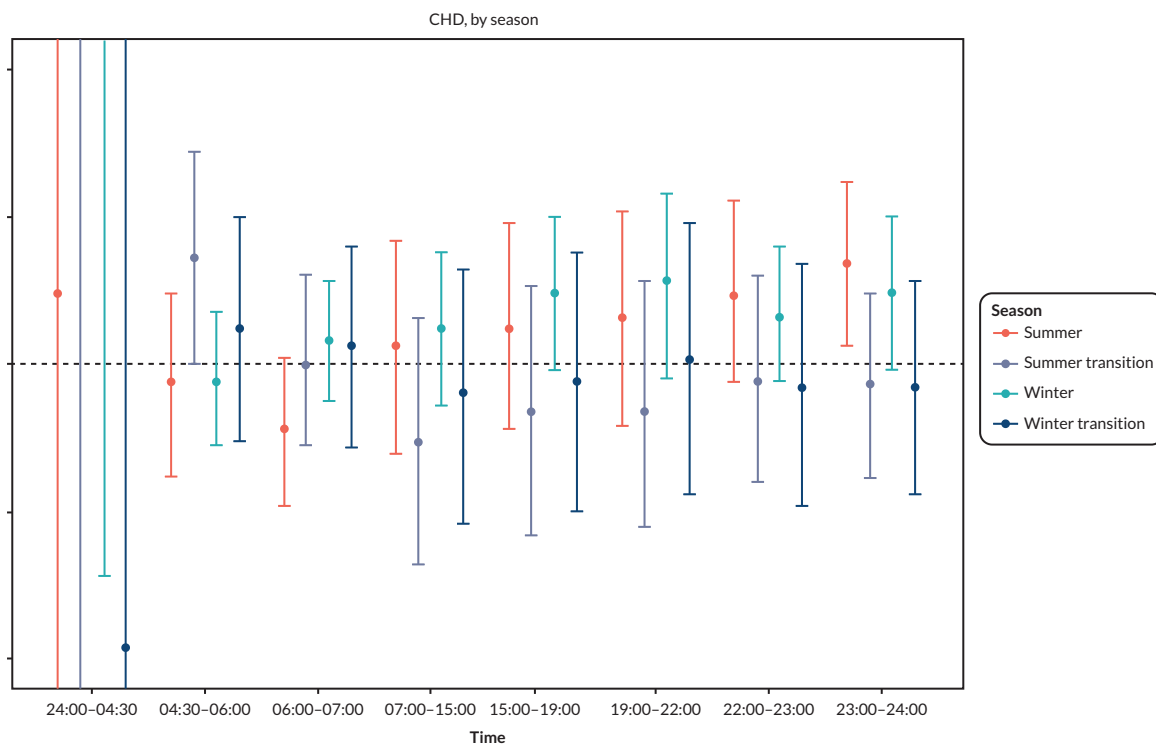


FIGURE 7 Sensitivity analyses: ORs and 95% CIs for hospital episodes due to all CVD, CHD and stroke per 5-dB increase LAeq at defined time points throughout the day, evening and night, including only the first hospital episode for the 60.8% of patients with more than one episode ($n = 269,915$). Estimates adjusted for $PM_{2.5}$ concentration, mean temperature and holiday effect.

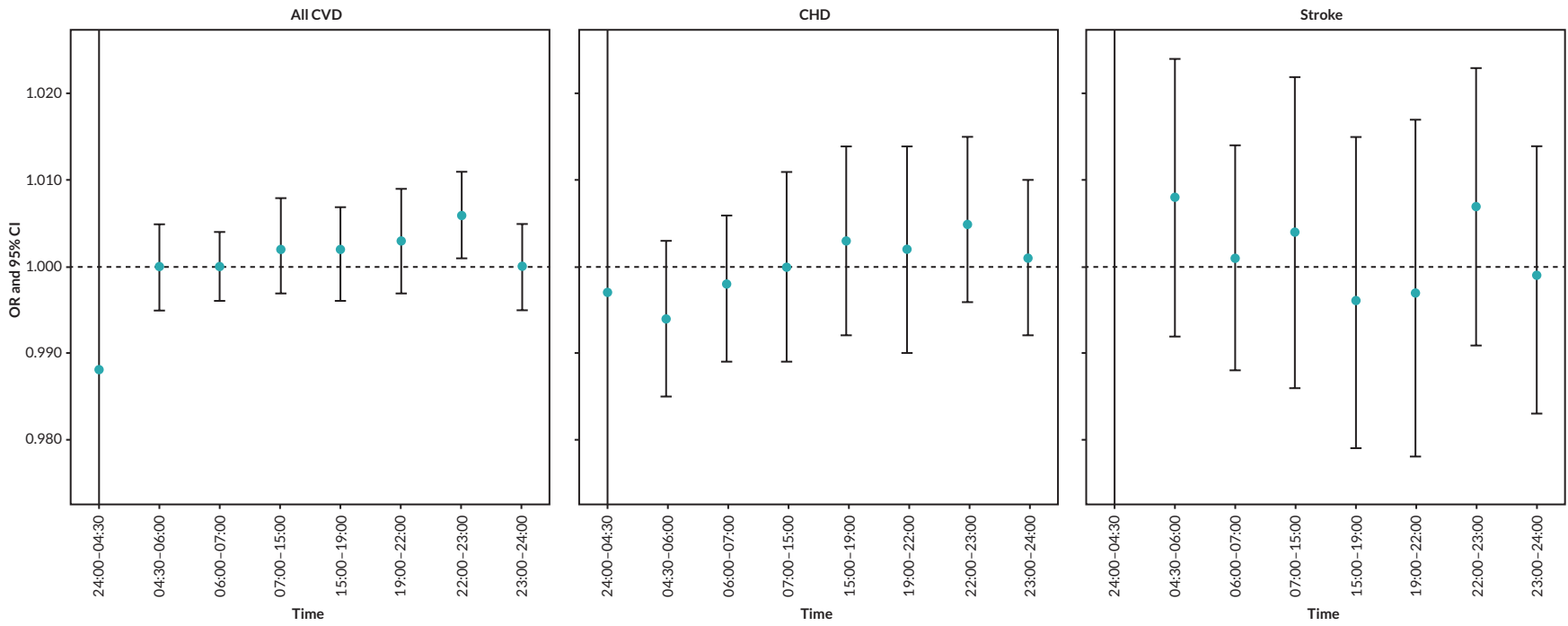


FIGURE 8 Odds ratios and 95% CIs for deaths due to all CVD, CHD and stroke per 5-dB increase LAeq at defined time points throughout the day, evening and night. Estimates adjusted for PM_{2.5} concentration, mean temperature and holiday effect.

TABLE 17 Odds ratio and 95% CIs for hospitalisations due to all CVD per 5-dB increase LAeq, stratified by low (\leq mean) or high ($>$ mean) CoV. Estimates adjusted for PM_{2.5} concentration, mean temperature and holiday effect

Group	Time band (hour)	OR	Lower limit	Upper limit
Low CoV ($<$ mean)	24.00–04.30	0.998	0.995	1.002
	04.30–06.00	1.001	0.997	1.006
	06.00–07.00	1.004	0.996	1.011
	07.00–15.00	1	0.99	1.011
	15.00–19.00	1.008	0.998	1.028
	19.00–22.00	1.012	1.002	1.023
	22.00–23.00	1.011	1.004	1.018
High CoV ($>$ mean)	23.00–24.00	1.005	1.001	1.008
	24.00–04.30	0.999	0.994	1.004
	04.30–06.00	0.997	0.995	0.999
	06.00–07.00	0.998	0.995	1.002
	07.00–15.00	1.002	0.997	1.006
	15.00–19.00	1.002	0.997	1.006
	19.00–22.00	1.002	0.996	1.007
22.00–23.00	1.004	0.999	1.009	
23.00–24.00	1.002	0.998	1.005	

Note

Bold values are statistically significant at the 95% level.

TABLE 18 Mean, median and SD noise levels for CVD hospital admissions in postcodes with low and high CoV

Period (hours)	Hospital admission			CoV
	Mean	Median	SD	
19.00–22.00	41.2	40.73	5.98	low
	42.59	42.21	7.72	high
22.00–23.00	41.02	40.32	6.61	low
	37.44	36.61	7.93	high
23.00–24.00	31.47	31.5	10.19	low
	24.02	23.68	10.31	high

Note

Bold values represent the mean, median, and standard deviation (SD) of noise levels for CVD hospital admissions in postcodes with low CoV.

Chapter 4 Discussion

Summary of results

Descriptive summary of daily aircraft noise data

We found that the morning shoulder period (06.00–07.00 hours) was the noisiest among all periods (mean 50.92 dB; 90th percentile 52.93 dB). Daytime (07.00–15.00 hours) aircraft noise levels were typically only slightly lower than the morning shoulder period (mean 49.87 dB; 90th percentile 51.50 dB). Night quota periods (23.30–04.30 hours) are typically times when people are sleeping, but we found that the average noise levels across postcodes from 23.00 to 24.00 hours and 24.00 to 04.30 hours were 41.06 and 29.81 dB, respectively.

We found that during 07.00–15.00 hours, postcodes within the study area experienced an average of eight flight events. Morning shoulder (06.00–07.00 hours) had an average of three events, while night-time (04.30–06.00 hours) had an average of one flight event.

Approaches to identifying respite and/or relief period

We did not have information on flight trials so attempted to identify relief periods (caused by e.g. wind direction changes as well as trials) in other ways. A priori definition of appreciable changes in noise did not identify enough postcodes. We therefore examined CoV of daily noise levels at postcodes. Highest variability was seen in night-time periods, with 24.00–04.30 hours having the highest mean CoV, followed by 04.30–06.00 and 23.00–24.00 hours.

Daily aircraft noise and material and health inequality

We examined the relationship between aircraft noise L_{day}, L_{eve} and L_{Aeq24} and three different deprivation/inequality measures. While postcodes near Heathrow with the least material or health deprivation experienced the lowest daily noise levels between 2014 and 2018, the relationship with deprivation measures, including the Carstairs index and fuel poverty, was complex and did not appear to have a clear gradient. A gradient was more evident between L_{night} and avoidable death rate. The fuel poverty rate had a weaker relationship with daily aircraft noise than Carstairs index and avoidable death rate.

Short-term impact of aircraft noise on cardiovascular morbidity and mortality

We included all recorded hospitalisations ($n = 442,442$) and deaths ($n = 49,443$) in 2014–18 due to CVD. We used conditional logistic regression to estimate the OR and adjusted for PM_{2.5} concentration, temperature and holidays. We estimated an increase in risk for a 5-dB increment in noise during the evening (L_{eve} OR = 1.005, 95% CI 1.000 to 1.010), particularly from 22.00 to 23.00 hours (OR = 1.006, 95% CI 1.002 to 1.010) for all CVD admissions. We found some evidence for effect modification by age, ethnicity, deprivation and season, but patterns were not consistent. The findings provide support for a potential mechanism through which aircraft noise may disturb sleep and elevate blood pressure, contributing to increased risk of cardiovascular hospitalisation.

Additionally, we found that an increased risk of CVD hospitalisation for increases in noise during the night-time hours (19.00–22.00, 22.00–23.00 and 23.00–24.00 hours) was only seen in postcodes with lower CoV. The average noise levels in the lower CoV postcodes were higher than other postcodes in two of the three periods – only giving partial support to the hypothesis that areas with higher noise levels and fewer relief periods have higher CVD admission risks. The impact of relief periods needs further research, ideally looking at other relevant outcomes (e.g. sleep disturbance, blood pressure, heart rate variability).

Reflections on what was and what was not successful in the programme

Delays

The dissemination of results was originally scheduled for 31 August 2020. The delays were caused by (1) relocation of the project from Imperial College to the University of Leicester, (2) complexities in the modelling process, which had not been anticipated by the noise consultancy conducting the modelling and (3) the COVID-19 pandemic, which among other impacts, slowed the recruitment of a researcher to conduct the analyses. The NIHR was supportive and responsive to communications about the delays throughout the project and a no-cost extension was agreed.

Strengths

We created and analysed a large dataset, which included daily average noise levels for approximately 155,000 postcodes near Heathrow airport. The AEDT version 3b was used to assess aircraft noise levels at each of the 155,000 postcodes in the vicinity of Heathrow airport from 1 January 2014 to 31 December 2018. These data are probably those with the highest resolution regarding Heathrow airport's daily aircraft noise levels. Our analysis of noise data concludes that exposure to daily aircraft noise remains an environmental problem for some communities near Heathrow airport.

The study area encompassed around 6.3 million people who resided near Heathrow airport in 2011. The high temporal resolution of our daily aircraft noise exposure data was an additional strength. Thirdly, our noise data were estimated for eight time bands, allowing us to distinguish between different daytime and night-time periods.

In the subsection where we examined the association between daily aircraft noise exposure and deprivation, we used a range of deprivation domains, including not only poverty but also health disparities.

Our study focused on the impact of short-term aircraft noise on cardiovascular morbidity and mortality and, to our knowledge, is one of only three such studies to date. In our analysis of population health, we included nearly all hospitalisations and deaths attributable to CVD, providing sufficient statistical power to detect an effect. By design, the case-crossover design controlled for significant measured and unmeasured confounders, such as lifestyle factors, ethnicity and age. Differentiating the effects of noise during specific times of the day, evening and night provided evidence for certain biological mechanisms observed in previous research.

Patient and public involvement

The methods and results were discussed at meetings of the study scientific advisory board, which includes representatives from the non-governmental Aviation Environment Federation and the Heathrow Association for the Control of Aircraft Noise, a community group set up over 50 years ago to represent people living under the Heathrow flight path.

Equality, diversity and inclusion issues

Our study included 442,442 naturally occurring hospital admissions and 49,443 fatalities from CVD, after removing some duplicate reports and records with no control variables. Males made up 58.0% of hospital admissions, with 56.7% being above the age of 65 years. Among those who stated their ethnicity, 9.4% were black and 11.4% were South Asian.

Limitations

Among the limitations are that the study area was designed to capture outer bounds of the CAA annual average aircraft noise contours in 2011. Some postcodes outside the study area could still be affected by aircraft noise but were not included in the analysis. However, given the spatial and temporal resolution and size of the data, this was a reasonable compromise.

In our analysis of deprivation, we used area-level not individual-level noise and deprivation estimates therefore the ecological fallacy may apply. Heathrow airport is situated close to highly populated areas, some of which are very wealthy, so may not be representative of other airports.

We were only able to examine outdoor noise levels and were not able to take account of housing characteristics, including double glazing, which may have affected indoor noise exposures.

Misclassification bias may also have been introduced because we used noise exposure at small geographical level rather than individual level. Another source of potential misclassification is that individuals may move outside of the postcode to which their exposure has been assigned at different periods throughout the day. We expect less exposure misclassification in the evening and night-time hours because individuals are more likely to be at their postcode of residence during these times. We also expect less misclassification among older individuals throughout the day and night, as they are less likely to travel away from home for work or school during the day. This may partially explain why effect estimates were highest during evening and night-time hours, and among individuals over the age of 65 years. Lastly, exposure misclassification may have been introduced because data on exact time of admission and death were not available, and we were therefore unable to define the precise window of exposure before an event occurs. To compensate for this, we used the average of lag 0 and lag 1 before the event day to ensure the defined exposure window captures the true exposure. Lastly, misclassification bias may be introduced due to moving home; according to English Housing Survey, between 9% and 11.3% of households in England moved home per year between 2011 and 2018.⁴⁷

Chapter 5 Conclusion

We have produced one of the most detailed datasets available on aircraft noise exposure at one of the world's busiest airports, covering a 5-year period at different times of day and night corresponding to airline operation periods. We conducted one of extremely few studies on short-term aircraft noise exposure and CVD hospitalisation and mortality. Our findings suggested that short-term noise exposure during the evening and night-time may be linked to an increased risk of all CVD hospital admissions, which would fit with a role for sleep disturbance. Numbers of deaths were small relative to hospital admissions, and we lacked statistical power to detect associations. Our results also suggest that sustained (low variability) noise exposure may be an important characteristic in relation to the health associations. This is partly supported by evidence that showed a significant association between aircraft noise during night-time and evening and hospital admissions in regions with low noise variability, but this was not consistent as to whether absolute noise levels were higher or lower.

In further analyses, we found some degree of correlation between different measures of deprivation and exposure to aircraft noise. While these were not straightforwardly linear, they did suggest that deprived areas were more exposed, particularly to night-time noise. In health analyses, there were interactions seen with deprivation (Carstairs index) with increase in risk of CVD hospitalisation in relation to acute aircraft noise exposure during late night hours among individuals residing in the third and fourth quintiles of deprivation (but not the most deprived quintile). This analysis is one of very few to investigate the relationship between aircraft noise and health inequality.

This information can be informative for national health policy, local residents and for airports and help inform future health and exposure studies. As air transport increases post-pandemic, information on noise exposures as well as views from community groups can inform future airport policies.

Recommendations for future research

We recommend the following future studies to further advance the knowledge:

- Make use of natural experiments to assess the impact of intervention on short-term aircraft noise exposure on CVD.
- Studies looking at interventions such as double glazing, other noise insulation or flight changes, linking indoor and outdoor noise measurements with objective measures of CVD risk; for example, blood pressure, heart rate variability, sleep and blood biomarkers.
- Further investigation of the relationship between noise variation and the risk of CVD, using novel noise metrics.
- Consideration of deprivation and ethnicity in all studies examining associations between aircraft noise and health outcomes as these had effect-modifying impacts in the current study related to cardiovascular health.

Implications for future studies

- We discovered that the period 24.00–04.30 hours had the greatest daily noise variation. Future studies on the health effects of aircraft noise pollution focused on Heathrow airport are advised to take the variation of night-time noise levels into account. Further work on noisy events is important (we only calculated these for N60 and N65 for 1 year and exploration of lower cut-points and other years would be helpful).
- We found that the morning quota period (06.00–07.00 hours) had the highest mean daily aircraft noise level per postcode, which contrasts sharply with the mean noise levels during the night

CONCLUSION

quota period (23.30–06.00 hours). This implies that future epidemiological studies focusing on Heathrow airport are recommended to distinguish this period from the regular night-time period (23.00–06.00 hours).

- We identified a positive link between avoidable death rate and daily aircraft noise exposure, particularly at night. This suggests a link between health deprivation and aircraft noise levels. Future epidemiological studies focusing on Heathrow airport are recommended to consider health deprivation as a potential confounder.

Additional information

Contributions of authors

Xiangpu Gong (<https://orcid.org/0000-0002-8985-9756>) (Research Associate, Epidemiology) helped to design noise exposure inequalities analyses, conducted the descriptive and exposure inequalities analyses and prepared results for publication.

Nicole Itzkowitz (<https://orcid.org/0000-0001-5076-3522>) (Research Assistant, Statistics) conducted the health analyses and prepared results for publication.

Calvin Jephcote (<https://orcid.org/0000-0003-0464-8448>) (Lecturer in Practice and Research Fellow in Geographical Information Science, Exposure Science) developed and collated environmental exposure assessment data, interpreted and prepared results for publication.

Kathryn Adams (<https://orcid.org/0000-0002-5305-709X>) (Research Assistant, Exposure Science) developed and collated environmental exposure assessment data.

Glory O Atilola (<https://orcid.org/0000-0002-4678-2015>) (Research Associate, Biostatistics) conducted the health analyses and prepared results for publication.

John Gulliver (<https://orcid.org/0000-0003-3423-2013>) (Professor of Environmental and Exposure Sciences) designed the study environmental exposure assessments, set up the contract with environmental consultants to provide the aircraft noise data, interpreted and prepared results for publication.

Marta Blangiardo (<https://orcid.org/0000-0002-1621-704X>) (Professor of Biostatistics) designed the study, designed the statistical analyses, conducted the health analyses, interpreted and prepared results for publication.

Anna Hansell (<https://orcid.org/0000-0001-9904-7447>) (Professor of Environmental Epidemiology, University of Leicester), initiated the study, conceived and designed the analyses, helped arrange ethics and governance permissions for health data access, set up contract specification with environmental consultants to provide the aircraft noise data, set up the study scientific advisory board, interpreted and prepared the results for publication.

Disclosure of interests

Full disclosure of interests: Completed ICMJE forms for all authors, including all related interests, are available in the toolkit on the NIHR Journals Library report publication page at <https://doi.org/10.3310/UTCE9104>.

Primary conflicts of interest: Conflicts of interest were reported by Nicole Itzkowitz (NIHR award), John Gulliver (NIHR awards), Anna Hansell (NIHR awards, chair of the Committee on the Medical Effects of Air Pollution, payment for role as PhD examiner in Finland 2022, Travel and subsistence costs for international 2-day meeting in Milan on air pollution).

Patient data statement

This work uses data provided by patients and collected by the NHS as part of their care and support. Using patient data is vital to improve health and care for everyone. There is huge potential to make better use of information from people's patient records, to understand more about disease, develop new treatments, monitor safety, and plan NHS services. Patient data should be kept safe and secure, to protect everyone's privacy, and it is important that there are safeguards to make sure that it is stored and used responsibly. Everyone should be able to find out about how patient data is used (#datasaveslives). You can find out more about the background to this citation here: <https://understandingpatientdata.org.uk/data-citation>.

Data-sharing statement

The aircraft noise exposure data are available to other academic researchers on request to the corresponding author.

Ethics statement

The study was covered by national research ethics approval from the London – South East Research Ethics Committee, reference 17/LO/0846 (date of opinion, 29 June 2017). Data access to confidential patient information without consent was covered by the Health Research Authority Confidentiality Advisory Group under Regulation 5 of the Health Service (Control of Patient Information) Regulations 2002 (section 251 support), reference: 20/CAG/0028 (outcome date 24 March 2020, section 251 Register Index Sheet application number A02476).

Carstairs index of deprivation (England and Wales, 2011)

The Carstairs index is an indicator of relative deprivation that is commonly used in spatial epidemiology to identify socioeconomic confounding.^{48,49} This deprivation index is constructed from four unweighted UK Census variables, which describe the level of male unemployment, overcrowding, private vehicle ownership and social composition in each community.

A revised form of the Carstairs index was constructed for census output area and lower layer super output areas in England and Wales, using the 2001 classification of low social classes devised by Norman.^{48,49} The revised low social class variable approximates its counterpart from the 1991 Census, developed to account for Office for National Statistics methodology and classification changes in later censuses.

Datasets from the 2011 Census were obtained from Nomis (www.nomisweb.co.uk), the official online delivery service of labour market statistics provided by the Office for National Statistics. Tables KS602EW, QS409EW, KS404EW and QS607EW contained the necessary information to create the Carstairs index for 2011 across England and Wales.

Table 19 lists the variable names, descriptions and formulas.

Each of these variables were z scored (mean-centred and divided by their SD) and all four z scores were summed to return an index value measuring the relative level of deprivation in each community. A value of 0 identifies communities that follow the national average of England and Wales, with negative values identifying increased affluence, and positive values identifying increased levels of deprivation.

TABLE 19 Data and formulas to calculate Carstairs index

Variable name	Description	Calculation
Proportion of 'male unemployment' (KS602EW)	'Unemployed males age 16–74 years' ÷ 'economically active males age 16–74 years'	$KS602EW0005 \div (KS602EW0002 + KS602EW0003 + KS602EW0004 + KS602EW0005 + KS602EW0006)$
Proportion of 'over-crowded households' (QS409EW)	('Over 1 and up to 1.5 persons per room' + 'over 1.5 persons per room') ÷ 'all households'	$(QS409EW0004 + QS409EW0005) \div QS409EW0001$
Proportion of 'households without vehicle ownership' (KS404EW)	'No cars or vans in household' ÷ 'all households'	$KS404EW0002 \div KS404EW0001$
Proportion of 'persons from a low social class' (QS607EW)	(L11.2 + L12.2 + L12.4 + L12.5 + L12.7 + L13.1 + L13.2 + L13.4 + L13.5) ÷ 'all persons'	$(QS607EW0035 + QS607EW0038 + QS607EW0040 + QS607EW0041 + QS607EW0043 + QS607EW0045 + QS607EW0046 + QS607EW0048 + QS607EW0049) \div QS607EW0001$

Carstairs index values for the 181,408 COA communities in England and Wales were then categorised in quintile groups, to control for any outliers within the data. This procedure was repeated for the 34,753 LSOA communities in England and Wales.

Nitrogen Dioxide Concentration

Nitrogen dioxide (NO₂) concentrations from all sources of pollution in µg/m³ were obtained from a land use regression-modelled raster surface with a 50-m grid resolution, developed by Gulliver *et al.*⁵⁰ This pollution model is based on a multivariate regression equation that describes the relationship between sample locations and environmental variables (i.e. rural–urban land cover classifications and road network data within several proximity buffer zones).

Postcode centroids (i.e. points representing the population-weighted centre of each postcode unit) were intersected with the continuous NO₂ raster surface, to extract an exposure value at each location.

Road transport noise (annual, 2013)

Modelled road-transport noise estimates were calculated in accordance with the CNOSSOS-EU common framework for noise assessment methods developed by the European Commission (2002/49/EC). For the purposes of this study, the CNOSSOS-EU model algorithms were implemented in PostgreSQL via the PostGIS v2.1 extension, following the protocol described by Morley *et al.*⁵¹

Annual average daily traffic counts and traffic speeds across the UK road network in 2013 enter the model, together with information relating to the surface roughness of land cover, building heights, wind profiles and average temperatures in 2013.

The CNOSSOS-EU model ran on the ALICE high-performance computing facility at the University of Leicester. [Figure 9](#) (adapted from Gulliver *et al.*⁵²), describes the workflow of the CNOSSOS-EU model.

In brief, the coordinates of each receptor (residential postcode centroid) are assigned to the closest building. The building façade that is likely to experience the most noise levels is identified (i.e. traffic count on a nearby road/road distance), and a receptor (point) is placed 1 m away from the building. A geographical information systems operation locates all major roads within a 1000-m radius and all minor roads within a 100-m radius of each receptor.

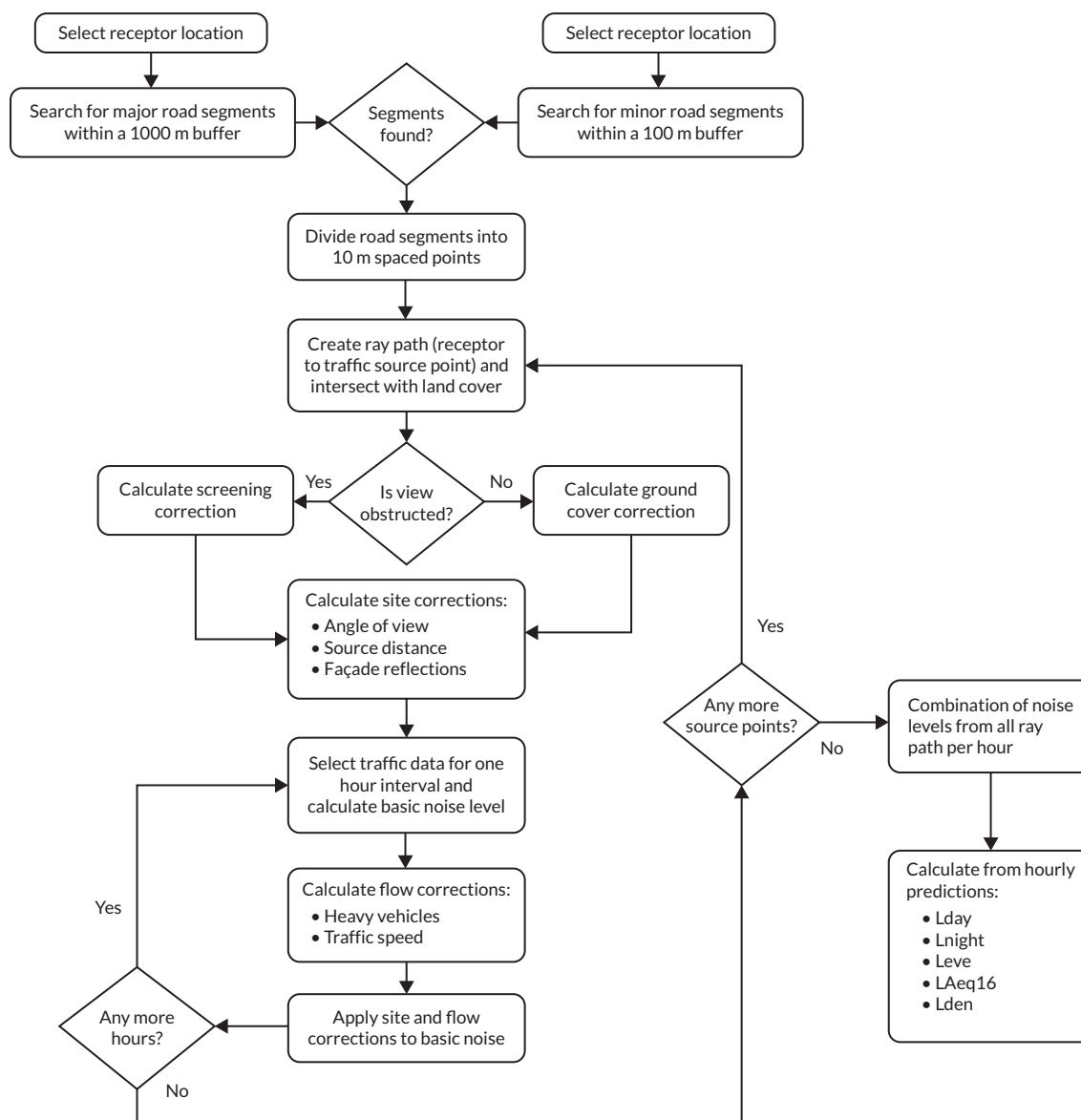


FIGURE 9 Workflow of the CNOSSOS-EU model.

Noise exposure from road transport was reported in accordance with five noise metrics that are ‘A’ frequency weighted. The ‘A’ weighting is a standard weighting of the audible frequencies designed to reflect the response of the human ear to noise (between 500 Hz and 6 kHz):

- **Lday** is the A-weighted equivalent noise level (Leq) over the 12-hour day period (07.00–19.00 hours), also known as the day noise indicator.
- **Levening** is the A-weighted equivalent noise level (Leq) over the 4-hour evening period (19.00–23.00 hours), also known as evening noise indicator.
- **Lnight** is the A-weighted equivalent noise level (Leq) over the 8-hour night period (23:00–07:00 hours), also known as the night noise indicator.
- **LAeq16** is the A-weighted equivalent noise level (Leq) over the 16-hour day period (07.00–23:00 hours).
- **Lden** is the A-weighted equivalent noise level (Leq) over a whole day, but with a penalty of +10 dB(A) for night-time noise (22.00–07.00 hours) and +5 dB(A) for evening noise (19.00–23.00 hours), also known as the day-evening-night noise indicator.

Hospitalisation and death registration data are held by and available through NHS Digital to researchers meeting relevant governance standards (<https://digital.nhs.uk/services/data-access-request-service-dars>).

All other data are available online. The avoidable death rate can be obtained from government statistics (gov.uk). The Carstairs index is accessible via the UK Data Archive. The Office for National Statistics provides the fuel poverty rate and percentage of non-white population per local authority district.

Information governance statement

The study was covered by national research ethics approval from the London – South East Research Ethics Committee, reference 17/LO/0846 (date of opinion, 29 June 2017). Data access to confidential patient information without consent was covered by the Health Research Authority Confidentiality Advisory Group under Regulation 5 of the Health Service (Control of Patient Information) Regulations 2002 (section 251 support) – HRA CAG reference: 20/CAG/0028 (CAG outcome date 24 March 2020, section 251 Register Index Sheet Application number A02476).

The University of Leicester is committed to handling all personal information in line with the UK Data Protection Act (2018) and the General Data Protection Regulation 2016/679. The University of Leicester acted as data controller and both University of Leicester and Imperial College London acted as data processors (the University of Leicester held the exposure data and made decisions on the analyses). Imperial College London held the health data and we supplied them with the exposure data, which was integrated into the health data, and joint decisions were taken on the health analyses).

For organisations where the sponsoring body is also the data controller under the data protection legislation, the University of Leicester is the data controller, and you can find out more about how we handle personal data, including how to exercise your individual rights and the contact details for our data protection officer here: <https://le.ac.uk/policies/insurance/gdpr-notice>.

Department of Health and Social Care disclaimer

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This monograph was published based on current knowledge at the time and date of publication. NIHR is committed to being inclusive and will continually monitor best practice and guidance in relation to terminology and language to ensure that we remain relevant to our stakeholders.

Publications

Gong X, Itzkowitz N, Atilola G, Adams K, Jephcote C, Blangiardo M, *et al*. The association between aircraft noise levels and material and health deprivation. Proceedings of the Inter-Noise conferences, 2022.

ADDITIONAL INFORMATION

Gong X, Adams K, Jephcote C, Itzkowitz N, Atilola G, Blangiardo M, *et al.* Characteristics of daily aircraft noise near Heathrow Airport. Proceedings of the 2022 UK and Ireland Exposure Science Meeting, 2022.

Itzkowitz N, Gong X, Adams K, Jephcote C, Gulliver J, Hansell A, Blangiardo M. Aircraft noise and cardiovascular morbidity and mortality near Heathrow Airport: a case-crossover study. Proceedings of the 34th Annual Conference of the International Society for Environmental Epidemiology, 2022.

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Appendix 1 Descriptive summary of daily aircraft noise levels by season

Season	Variables	Aircraft (N)	Mean	SD	Min	Max	90th percentile
Summer	LAeq 04.30–06.00	740,333	24.92	7.821	7.123	58.21	36.3
	LAeq 06.00–07.00	740,333	40.68	6.43	16.39	76.09	48.84
	LAeq 07.00–15.00	740,333	42.03	5.312	27.25	72.1	48.84
	LAeq 15.00–19.00	740,333	41.69	5.238	27.7	69.67	48.4
	LAeq 19.00–22.00	740,333	41.25	5.326	22.86	68.79	48.23
	LAeq 22.00–23.00	740,333	36.96	7.186	12.65	70.29	45.63
	LAeq 23.00–24.00	740,333	30.64	6.505	12.65	67.17	39.68
	LAeq 24.00–04.30	696,831	14.98	6.392	0.00	55.89	22.99
	N60 04.30–06.00	155,671	0.64	1.444	0	7.196	3.09
	N60 06.00–07.00	155,671	3.194	6.206	0	38.64	11.82
	N65 07.00–15.00	155,671	6.848	23.95	0	323.8	12.68
	N65 15.00–19.00	155,671	3.158	11.25	0	170.4	5.95
	N65 19.00–22.00	155,671	2.331	8.247	0	117.3	4.1
	N65 22.00–23.00	155,671	0.554	2.869	0	33.39	0.22
	N60 23.00–24.00	155,671	0.722	1.65	0	21.24	2.67
	N60 24.00–04.30	155,671	0.338	0.856	0	8	1
Summer transition	LAeq 04.30–06.00	740,333	23.71	6.262	10.03	54.02	32.49
	LAeq 06.00–07.00	740,333	39.88	5.662	19.47	74	46.7
	LAeq 07.00–15.00	740,333	41.65	5.658	26.46	75.11	48.41
	LAeq 15.00–19.00	740,333	41.63	5.388	27.44	74.51	48.1
	LAeq 19.00–22.00	740,333	41.65	5.239	24.9	75.17	48.04
	LAeq 22.00–23.00	740,333	39.94	5.111	20.97	69.87	46.54
	LAeq 23.00–24.00	740,333	30.2	6.079	12.17	63.72	38.42
	LAeq 24.00–04.30	740,333	14.22	5.147	1.344	44.01	20.9
	N60 04.30 –.600	115,206	0.579	1.263	0	5.951	2.49
	N60 06.00–07.00	115,206	2.912	5.503	0	32.62	11.44
	N65 07.00–15.00	115,206	7.738	24.23	0	320.9	16.47
	N65 15.00–19.00	115,206	3.549	11.57	0	163.5	7.01
	N65 19.00–22.00	115,206	2.599	8.395	0	115.5	5.03
	N65 22.00–23.00	115,206	0.796	2.306	0	32.57	2.19
	N60 23.00–24.00	115,206	0.411	0.78	0	8.897	1.21
	N60 24.00–04.30	115,206	0.0768	0.132	0	1.194	0.2

continued

APPENDIX 1

Season	Variables	Aircraft (N)	Mean	SD	Min	Max	90th percentile
Winter	LAeq 04.30–06.00	740,333	28.98	7.055	9.872	61.08	38.79
	LAeq 06.00–07.00	740,333	41.51	5.989	18.39	74.98	49.08
	LAeq 07.00–15.00	740,333	43.17	5.11	28.61	72.32	49.82
	LAeq 15.00–19.00	740,333	42.6	5.111	27.97	69.86	49.13
	LAeq 19.00–22.00	740,333	42.34	5.385	24.76	74.94	49.53
	LAeq 22.00–23.00	740,333	38.44	5.764	19.54	67.61	46.09
	LAeq 23.00–24.00	740,333	23.72	6.442	8.987	58.05	33.31
	LAeq 24.00–04.30	727,635	12.5	5.937	0.000267	51.31	19.93
	N60 04.30–06.00	115,206	0.911	1.943	0	8.742	3.97
	N60 06.00–07.00	115,206	2.833	5.324	0	30.74	11.18
	N65 07.00–15.00	115,206	8.216	23.46	0	300.4	18.89
	N65 15.00–19.00	115,206	3.873	11.67	0	158.8	7.88
	N65 19.00–2.200	115,206	2.675	7.985	0	108.4	5.55
	N65 22.00–23.00	115,206	0.629	1.676	0	23.24	1.77
	N60 23.00–2.400	115,206	0.21	0.379	0	4.384	0.58
	N60 24.00–04.30	115,206	0.107	0.192	0	2.065	0.29
Winter transition	LAeq 04.30–06.00	740,333	24.15	8.307	5.009	64.36	35.13
	LAeq 06.00–07.00	740,333	41.35	6.626	13.86	78.37	49.66
	LAeq 07.00–15.00	740,333	42.64	5.445	27.37	78.34	49.32
	LAeq 15.00–19.00	740,333	42.33	5.466	26.41	75.24	49.3
	LAeq 19.00–22.00	740,333	42.03	5.548	21.68	75.46	48.9
	LAeq 22.00–23.00	740,333	40.43	5.805	20.43	75.46	48.15
	LAeq 23.00–24.00	740,333	31.11	7.531	11.91	72.29	41.34
	LAeq 24.00–04.30	702,071	14.12	5.69	7.16E-05	53.42	21.51

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