Seasonal Variations in the Synoptic Climatology of Air Pollution in Birmingham, United Kingdom

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Abstract

A synoptic typing approach was undertaken to examine the seasonal relationship (winter versus summer) between air mass types and pollutant concentrations of O₃, PM₁₀, NO_x, NO₂ and CO in Birmingham, United Kingdom from 2000 to 2015. Daily means of seven surface meteorological variables were entered into a P-mode principal component analysis. Three principal components explained 72.2% (72.9%) of the variance in winter (summer). Cluster analysis was used to group together days with similar PC scores and thus homogeneous meteorological conditions. Six clusters provided the best air mass classification in both seasons. High pollutant concentrations were associated with anticyclonic types. In particular, tropical (polar) continental air mass type was most likely to produce extremely high concentrations in summer (winter). In winter, a sequence of Polar Continental (cool and humid) and Binary Mid-latitude Anticyclonic Maritime – Sub-Polar Cyclonic Maritime (cold and dry) induced severe pollution episodes in all pollutants. Whilst the mean duration of severe pollution episodes varied little between winter and summer (O₃ was an exception, with severe episodes lasting 20% longer in summer), high pollutant extremes were more common in winter. This was due to more favourable meteorological conditions (e.g., temperature inversions) and increased anthropogenic emissions during the cold season.

1. Introduction

The Industrial Revolution brought considerable wealth and development to the coal-powered industrial towns of Britain, including the Black Country to the west of Birmingham and wider West Midlands area. Economic development came at a cost, with nitrogen oxides (NO_x), carbon monoxide (CO) and ozone (O₃) emissions in the Black Country being 50 times higher than present day concentrations (Hatton, 2017). The considerable epidemiological impacts led to the Clean Air Acts of 1956, 1968 and 1993 being passed to regulate domestic, commercial and industrial emissions (Abbott *et al.*, 2012). Birmingham is Britain's second largest city, with a population of 1.22 million. Multiple industrial sources, a major airport and congested roads result in high pollutant emissions. In particular, combustion from diesel engines accounts for the majority of the city's NO_x emissions. The Environment Act 1995 produced a national air quality strategy, which gave jurisdiction to local authorities such as Birmingham City Council to review and tackle their air pollution problem (DEFRA, 2018). In 2003, Birmingham was designated as an Air Quality Management Area (AQMA), in which pollutant concentrations must be regularly monitored and policies enacted to reduce levels of harmful pollutants such as NO₂ and PM₁₀. Despite this, the concentration of NO₂ in Birmingham is up to 50% higher than recommended levels (DEFRA and DfT, 2017). Bloss (2018) reported that poor air quality causes 900 premature deaths in Birmingham every year. In order to satisfy annual and winter pollutant targets, Birmingham City Council are introducing a £8 charge for older private cars to enter the city centre from June 2021.

Since the 1980s a growing body of research referred to as the 'Delaware Approach' (Kalkstein and Corrigan, 1986) has established that the physical and thermodynamic properties of the urban boundary layer play a primary role in influencing the dispersion of pollutants (McGregor and Bamzels, 1995; Shahgedanova *et al.*, 1998). This synoptic typing Delaware Approach adopts an air mass approach to understanding the relationship between weather and pollution. The advantage of such an approach is that it considers the synergistic behaviour of a multitude of atmospheric variables rather than analysing variables such as temperature and humidity in isolation (Tselepidaki *et al.*, 1995; McGregor, 1999). Crucially, the grouping of multiple surface weather variables in the air mass approach allows one to infer whether conditions are likely to favour pollutant accumulation or suppression. Principal component analysis (PCA) and cluster analysis (CA) are often used to derive the air mass types. PCA identifies groups of highly correlated variables over time. The resulting principal components are linearly independent and explain the maximum amount of variance possible (Yarnal, 1993). Many researchers then cluster the PC scores to identify the air mass types. Studies that have used the PCA and CA approach to synoptic typing include McGregor and Bamzels in Birmingham, UK (1995) and Shahgedanova *et al.* (1998) in Moscow. Cheng and Lam (2000) found that air mass types with low wind speeds, low temperatures and anticyclonic conditions were conducive to extremely high CO concentrations in Hong Kong. Kalkstein and Corrigan (1986) showed that maritime air masses - characterised by low pressure and high winds - resulted in low SO₂ concentrations in Delaware, US due to effective mixing and ventilation. Similarly, CO concentrations halved with maritime polar air masses in Moscow (Shahgedanova *et al.*, 1998).

Many studies have acknowledged the tendency of urban air pollutants to persist (Chen *et al.*, 2016). This persistence is a complex phenomenon governed by many inter-related factors. Chelani (2013) showed that CO persistence was dependent on preceding concentrations. Ozone persistence is not dependent on antecedent levels; but is primarily dependent on NO_x and CO emissions as precursor sources, and strong sunlight (Varostos *et al.*, 2003). Lonati *et al.* (2011) found that stable anticyclonic air mass types led to persistently high PM₁₀ concentrations in Milan, Italy. The implications of severe PM₁₀ episodes for daily hospital respiratory admissions in Birmingham were explored by McGregor *et al.* (1999). Xie *et al.* (2015) found that a 10 µgm⁻³ increase in PM_{2.5} was associated with a 0.27% increase in ischaemic heart disease mortality.

Despite the well-established nature of the weather-pollution relationship, seasonal differences in this association have received comparatively little attention. For instance, McGregor *et al.*'s (1999) application of the air mass type to Birmingham considered the winter only. A winter-only study might not be generalisable to other seasons, given the large seasonal variations that exist in weather conditions and pollutant emissions. Domestic energy use increases significantly in colder winter months, with people more likely to use their cars for shorter journeys.

This paper uses a synoptic typing approach to derive air mass types for both winter and summer in Birmingham, so enabling a comparative analysis of the meteorological controls on pollutants between the two seasons to be conducted. The objectives of this paper are as follows:

- (i) to derive winter and summer air mass types for Birmingham;
- (ii) to assess whether a relationship exists between the air mass types and mean and extreme pollutant concentrations; and
- (iii) to examine the seasonal differences in the above relationships

This paper is structured as follows. Section two describes the collection and analyses of data. Results are described in section three and then discussed in section four. Conclusions are drawn in section five.

2. Data And Methodology

Seven meteorological variables were obtained from the British Atmospheric Data Centre (BADC) for Coleshill (10km east of Birmingham City Centre): mean sea level pressure (MSLP), air temperature (TEMP), wind speed, wind direction, relative humidity (RH), cloud cover (CLOUD) and rainfall (RAIN). Hourly values of these variables were abstracted for winters (DJF) and summers (JJA) from 2000 to 2015, and used to calculate daily means. The sample sizes were 1,354 and 1,380 days for winter and summer respectively. Wind direction cannot be entered into the PCA because it is not on a linear scale. To solve this problem, wind speed and direction were used to calculate the zonal (U) and meridional (V) wind components, which were then entered into the PCA.

Daily mean concentrations of O₃, PM₁₀, NO_x, NO₂ and CO were abstracted from DEFRA's air quality archive. Unfortunately, data were not available for one site for the complete 2000-2015 period, so two sites had to be used: Birmingham Centre (2000-4) and Tyburn (2004-15). It was decided that Birmingham Tyburn (4.5 km north east of the City Centre) would make up the majority of the dataset given its closer proximity to the meteorological station at Coleshill. The distance between Coleshill and Tyburn (8 km) is acceptable here because air masses typically cover areas greater than hundreds of square kilometres (Kalkstein and Corrigan, 1986). The Tyburn readings are taken next to a busy arterial road and so are significantly higher than readings at Birmingham Centre. In order to produce a homogeneous time series, regression models were used to calibrate the 2000-4 values at Birmingham Centre to those at Tyburn (table 1). The regression equations were also used to infill any missing values at Tyburn from the other site. Calibration of CO was not performed because it was only measured at Birmingham Centre until 2008.

Daily means of MSLP, TEMP, U, V, RH, CLOUD and log₁₀RAIN (dry days were replaced with 0.01mm) were entered into separate P-mode PCAs for winter and summer. The correlation matrix was used because it standardises the variance of each variable to unity, thus eliminating the problem of different measurement scales. It is desirable for each variable to load strongly on to one principal component (PC). This is called simple structure. In the winter PCA, MSLP, TEMP and CLOUD were complex variables; and loaded strongly on to multiple PCs. An orthogonal Varimax rotation was thus applied to the winter PC axes in order to achieve better simple structure. No complex variables were identified for the summer PCA, which meant that rotation was unnecessary. PCs with eigenvalues greater than one were retained because these explain more variance than one original variable (Kaiser, 1960).

Cluster analysis was used to group together days with similar PC scores and thus homogeneous meteorological conditions. Again, separate cluster analyses were run for winter and summer using the group average method. Whilst Ward's method is most popular, it prioritises equal cluster sizes at the expense of physical air mass interpretation by grouping extreme days with non-extreme days (Kalkstein *et al.*, 1987). By contrast, the group average method focuses on 100% likeness of adjoining cluster cases by prioritising minimisation of within group variance and maximisation of between group variance (Shahgedanova *et al.*, 1998). The squared Euclidean distance was used as the similarity measure. To determine the number of clusters in both seasons, a dendrogram and scree-plotted agglomeration schedule were used, the latter of which plots the co-efficient of fusion against the number of clusters (Gower and Barnfield, 1975). These methods allow the researcher to deduce significant between-cluster heterogeneity and thus unique air mass types.

An independent samples one-way analysis of variance (ANOVA) was conducted for winter and summer to investigate whether pollutant concentrations vary between the air mass types. Logged values of NO_x, PM₁₀ and CO were utilised to satisfy the normality assumption of ANOVA. Whilst ANOVA reveals whether the air masses are different as a whole, it does not show whether one air mass shows a statistically significant difference from the other air masses. As such, a Tukey test was performed, in which significant differences in pollutant concentrations between pairs of air masses were identified.

For each pollutant and for both summer and winter datasets separately, a Chi-Square test was used to determine whether the most polluted days (standardised scores > 2) are disproportionately associated with certain air mass types. The null hypothesis (H₀) of the test assumes that the number of most polluted days is proportional to the frequency of occurrence of that air mass type. For example, if an air mass occurred on 20% of days throughout the entire data set, then H₀ assumes that this air mass accounts for 20% of the most polluted days. The Chi-Square test was also used to ascertain whether three thresholds (high, very high and exceptionally high; see table 2) for poor air quality were more likely to be exceeded in summer or winter.

A severe pollution episode was defined as a sequence of at least four days with concentrations on all days above the 90th percentile. The dominant air mass type for each episode was noted.

3. Results

Principal component analysis

For the winter, three PCs explain 72.2% of the variance. PC 1 (explained variance = 26.0%) has a strong positive association with log₁₀RAIN and V; and a strong negative relationship with MSLP (table 3). PC 1 captures the dominant mode of variability in winter weather conditions, which tend either to be mild and wet with low pressure (high positive PC 1 scores) or cold and dry with high pressure (high negative PC 1 scores). RH and CLOUD have high positive loadings on PC 2 (explained variance = 24.5%); U is the only variable with a strong negative PC 2 loading, which suggests that increased RH and CLOUD are associated with easterly flows. This is logical because easterly winds are cold in winter, which results in a higher RH and often overcast conditions. By contrast, westerly winds are often accompanied with clearer skies. PC 2 can thus be interpreted as the state of the sky. PC 3 (explained variance = 21.8%) is a thermal component. It has a strong positive relationship with TEMP and CLOUD; and a weaker positive relationship with V, U and log₁₀RAIN. On days with high positive PC 3 scores, temperatures are higher than average, with moderate south-westerlies, extensive cloud cover and drizzle. Days with high negative PC 3 scores are much cooler and clearer.

Three PCs explain 72.9% of the variance during the summer. PC 1 (explained variance = 39.2%) has a strong positive relationship with $\log_{10}$ RAIN, RH and CLOUD; and a strong negative association with MSLP. On days with high positive PC1 scores, low pressure areas will be present over Britain, bringing cloud and significant rainfall. Days with high negative PC 1 scores are characterised by anticyclonic conditions, with warmer temperatures and blue skies. PC 1 is thus a hygrometric and pressure component. PC 2 (explained variance = 18%) is capturing the strong co-variability between south-easterly airflows from Continental Europe and warmer temperatures. PC 3 (explained variance = 15.7%) has a strong positive association with U, and a moderate inverse association with RH and MSLP. Days with high positive PC 3 scores have strong westerly winds and moderately low pressure, in which depressions driven by strong Atlantic westerlies travel eastwards over Britain. On days with high negative PC 3 scores, the wind tends to blow from the east and MSLP is higher. PC 3 is thus a zonal wind component.

Winter air mass types

For the winter, the dendrogram and agglomeration schedule indicate that the retention of six clusters is most appropriate. The properties of each air mass type are now described (figure 1 and table 4).

Type 1: Anticyclonic Tropical Maritime

With south-westerly winds, conditions are relatively mild with high cloud cover.

Above average pressure suppresses the amount of rainfall. Typically associated with the passage of warm fronts, although cold and occluded fronts can on occasions bring heavier rainfall to the English Midlands. High pressure is positioned over France and Iberia, with depressions moving eastwards between Scotland and Iceland.

Type 2: Unstable Tropical Maritime

The wettest cluster with the lowest pressure. Birmingham is affected by deep Atlantic lows, cold fronts and strong south-westerly winds.

Type 3: Polar Continental

Much cooler than the first two air mass types, and so has a higher relative humidity. The wind direction is variable, but always from a cold, continental source. Pressure is usually high; rainfall is moderate, with occasional heavy downpours. Nocturnal radiative cooling brings morning fog, which dissipates by early afternoon.

Type 4: Binary mid-latitude anticyclonic maritime – sub-polar cyclonic maritime (AM-SCM) (McGregor, 1999)

Northerly winds drive very cold conditions, with low relative humidity and cloud cover. Typically features a well-developed mid-Atlantic anticyclone, with a depression of moderate intensity lying between Iceland and Scandinavia. Rainfall decreases as the mid-Atlantic high moves towards Britain.

Type 5: Extreme Arctic Maritime

Occurs on only three days. Characterised by north-westerly winds, which bring bitterly cold temperatures.

Type 6: Returning Polar Maritime

This air mass is characterised by fairly mild temperatures, pressure close to the long-term average, strong westerly winds, and moderate cloud cover and rainfall. Depressions track eastwards between Iceland and Norway. When the isobars are traced back into the western Atlantic, they originate from the Greenland and Canada region.

Summer air mass types

Six clusters were again considered to provide the best air mass classification. The properties of each air mass type are now described (figure 2 and table 5).

Type 1: Tropical Maritime

Predominantly wet and cloudy, with moderate temperatures and strong south-westerlies. This air mass typically originates from the Azores and contains active frontal systems. Rainfall was as high as 33.6mm on one day.

Type 2: Transitional Polar Arctic

Low pressure centred over southern England or northern France leads to cold north-easterly winds transporting Scandinavian or Arctic air to Birmingham. This is by far the wettest air mass type. Anticyclones in the mid-Atlantic and Eastern Europe are common in this situation.

Type 3: Stable Anticyclonic Tropical

Characterised by moderate westerlies, temperatures and cloud cover. Whilst the air has a long sea track, high pressure over southern Britain is the dominant feature and results in mainly dry conditions.

Type 4: Mild Stable Polar Continental

High pressure and north-easterly winds usually bring low rainfall. Significant low-level cloud cover brings morning fog, with potential for heavy rain when there are disturbances embedded in the flow and a longer North Sea fetch.

Type 5: Tropical Continental

The driest and warmest air mass type, with low cloud cover and a gentle south-easterly breeze. High pressure centred in the southern North Sea leads to settled conditions.

Type 6: Polar Maritime

This cluster contains only 16 days. Temperatures are on the cool side, with light to moderate rainfall. A low is typically situated over north-west Scotland, which brings strong westerlies to Britain.

Pollutant variability with air mass type

All five pollutants show statistically significant differences between the six air mass types at the 99.9% confidence level in both winter and summer (table 6). As Snedecor's variance ratio F is considerably greater than one, there is far more variance in pollutant concentrations between the air mass types than within types. For all five pollutants, the F ratio is larger in the winter. This means that concentrations vary more with air mass type in winter than in summer.

Difference matrices are used to represent the results of the post-hoc Tukey tests. In these matrices, matrix cell entry denotes a statistically significant difference in the concentration of the given pollutant at the 99% confidence level.

In the winter (figure 3), NO₂ and O₃ have the most significant pairings with 12 and 10 respectively whilst PM10 has the fewest at six. Returning Polar Maritime (air mass type six) is responsible for the largest number of significant pairings (20), with Extreme Arctic Maritime (type five) having the least pairings (eight).

During the summer (figure 4), PM10 and CO have the largest and smallest number of statistically significant pairings at 11 and six respectively. The Transitional Polar Arctic (type two), Stable Dry Anticyclonic Tropical (type three) and Tropical Continental (type five) all have the largest number of significant pairings (16). This means that these three air masses are the most distinctive in terms of concentrations. In particular, air mass type three shows statistically significant differences at the 99% confidence level with types two and four across all five pollutants.

Distribution of air mass types on the most polluted days

Winter air mass type five was excluded from the Chi-Square test because it occurred on only three days and so violated the 20% rule for expected counts. The test results show that the most polluted days in winter (Z score > 2) show a significant association with the air mass type at the 99.9% confidence level (table 7 and figure 5). All pollutants show a statistically significant relationship with summer air mass type at the 0.01 level or better (table 7). In particular, high NO_x, PM10 and CO concentrations occur disproportionately under summer air mass type three, accounting for 49%, 51% and 65% of high concentration days respectively (figure 6). Air mass type six is not present for any pollutant except O₃.

Severe pollution episodes

A severe pollution episode was defined as a sequence of at least four days with concentrations above the 90th percentile. Figure 7 clearly shows that Polar Continental air mass type (AT) 3 and AM-SCM AT 4 are the most dominant air masses in severe pollution episodes in winter. Extreme Arctic Maritime (AT 5) and Returning Polar Maritime (AT 6) do not appear at all. Anticyclonic Tropical Maritime (AT 1) and AM-SCM (AT 4) often persist for the majority of the severe pollution episodes in which they dominate.

In summer, Tropical Continental AT 5 and Mild Stable Polar Continental AT 4 are the air masses most conducive to severe pollution episodes (figure 8). Polar Maritime AT 6 does not appear at all, with Transitional Polar Arctic AT 2 also making minimal impact. Tropical Continental (AT 5) and Stable Anticyclonic Tropical (AT 3) tend to persist for the majority of the severe pollution episode in which they dominate.

Cross-seasonal comparison of pollution episode severity

Figure 9 reveals that whilst the difference in length in days of pollution episodes varies only minimally between summer and winter for each pollutant, average concentrations for all pollutants are higher in the winter. The exception is O_3 , which has greater concentrations and longer episode duration in summer. This is to be expected given that increased summer sunshine hours are conducive to increased photolysis. NO_x has much higher concentrations in winter than summer, whilst PM10, CO and NO_2 are only slightly higher.

Cross-seasonal comparison of pollution extreme values

The thresholds for high, very high and exceptionally high concentrations are defined in table 2. The Chi-Square test is applied to determine whether these thresholds are more likely to be exceeded in winter or summer. Apart from O_3 ($p = 0.024$), the pollutants all show a statistically significant difference in their frequency of occurrence between winter and summer at the 99.9% confidence level (figure 10). These high pollutant thresholds are far more likely to be exceeded in winter than in summer, with the exception of O_3 . There were 317 days in winter when NO_2 exceeded $50 \mu\text{gm}^{-3}$ but only 18 such days in summer. Similarly, PM10 exceeded $50 \mu\text{gm}^{-3}$ on 37 winter days but only six summer days.

4. Discussion

Principal component analysis

In winter, days with high positive PC 1 scores (explained variance = 26.0%) are characterised by unstable, precipitation-bearing Atlantic air mass passage over Britain. The winter-only assessment of synoptic airmass types in the same location by McGregor et al. (1999) shared this atmospheric pattern, although it accounted for only 12.5% of the variability as PC 3. PC 2 is a hygrometric zonal wind component, in which strong positive days induce cloudy, foggy conditions with cold easterly winds. Kalkstein and Corrigan's (1986) identified these conditions, in which north-easterly winds brought low-level mist and thus poor visibility to Delaware, whilst PC 5 in Moscow (Shahgedanova et al., 1998) was strongly related to cloud cover. PC 3 (21.8%) is a thermal component; conditions are anomalously mild on days with high positive scores. This pattern was also found in Birmingham by McGregor and Bamzeli (1995) and McGregor et al. (1999), but appeared as their PC 1 rather than PC 3. The statistically significant PCs collectively reveal that in winter the weather in Birmingham tends either to be mild and wet or cold and dry.

In summer, PC 1 (39.2%) is a hygrometric pressure component, with humid and cyclonic conditions occurring on days with high positive scores. This mode of variability was also identified by McGregor and Bamzeli (1995); and Shahgedanova et al. (1998), who identified summer PC 1 as describing pressure-humidity-stability variations. PC 2 (18%) is a thermal meridional wind component; hot, southerly winds occur on days with high positive scores. Kalkstein and Corrigan (1986) identified thermal variation as the most significant mode of variation (i.e., PC 1) in Wilmington, Delaware. PC 3 (15.7%) is a zonal wind component, with strong westerlies propagating Atlantic low pressure systems over Britain on days with high positive PC 3 scores. Significantly, McGregor and Bamzeli (1995) found Birmingham's PC 3 to be a zonal flow component.

Air mass types

Cluster Analysis (CA) identified six air mass types in winter and summer and winter. McGregor et al. (1999) also found six winter air mass types in Birmingham, including the Binary Mid-latitude Anticyclonic Maritime – Sub-polar Cyclonic Maritime (AM-SCM), that appeared as winter air mass type four in this study. McGregor's (1999) six winter Birmingham air mass types were similar to those in this study. In winter (summer), five (four) of the air mass types were maritime, which reflects the dominance of westerly winds in Britain's wind climatology.

Pollutant variability with air mass type

The ANOVA (table 6) and post-hoc Tukey tests (figure 3) showed statistically significant pollutant concentration differences between winter air mass types. These differences occur disproportionately when Returning Polar Maritime (air mass type six), which has the lowest average pollutant concentrations, is compared to the other five air mass types. The Returning Polar Maritime air mass is characterised by low pressure and strong upper-level westerlies that encourage the dispersion of pollutants. This was also the case in Moscow, where Shahgedanova et al. (1998) found that CO emissions halved under polar maritime air masses. Anticyclonic winter air mass types (AT), such as AT 4 AM-SCM, are conducive to pollutant stagnation. The northerly and easterly flows that occur in these air mass types are cold and stable. This accentuates ground-based temperature inversions, which suppress turbulence and the boundary layer mixing height (McGregor, 1999). The calm conditions suppress vertical pollutant diffusion in winter. Shahgedanova et al. (1998) also found that the highest CO concentrations in Moscow occurred in winter anticyclonic air mass types. Crucially, McGregor et al. (1999) identified identical cold, anticyclonic air mass types in Birmingham to those in this study (AT 3 and AT4); and found they produced the greatest conditional probabilities of inducing the highest respiratory admissions index class when PM10 was simultaneously the highest.

The ANOVA (table 6) and post-hoc Tukey test (figure 4) results showed statistically significant pollutant concentration differences between summer air mass types. These differences disproportionately occur when Tropical Continental (AT 5), which has the highest concentration for three of the five pollutants, is compared to the other five air mass types. Tropical continental air masses are typically characterised by anticyclonic conditions that are conducive to low cloud cover, low wind speeds and atmospheric stability (Comrie, 1992). This limits the ventilation and dispersion of pollutants because advective mixing is suppressed and a strong atmospheric subsidence exists, resulting in anomalously high pollutant concentrations. Cheng and Lam (2000) found that SO_2

concentrations in Hong Kong increased six-fold from cyclonic to anticyclonic air mass types; in this study, mean NO_x in Birmingham under the Tropical Continental air mass is almost double that of the Polar Maritime air mass type. The passage of the Tropical Continental air mass over Europe's industrial and urban centres is also conducive to high pollutant concentrations.

This study indicated a significant difference in O_3 concentrations between air mass types in both winter and summer. Winter Returning Polar Maritime (AT 6) and summer Tropical Continental (AT 5) possessed the highest mean concentrations of $55.0 \mu\text{g m}^{-3}$ and $60.7 \mu\text{g m}^{-3}$ respectively. These air mass types are characterised by low cloud cover. The resulting increase in radiation causes NO_2 to dissociate NO_2 into $\text{NO} + \text{O}$, which are the precursors of O_3 formation via photolysis (McGregor and Bamzels, 1995). This link can be corroborated by the low NO_2 concentrations that occur in air mass types associated with less cloud cover; e.g., winter AT 6 mean = $24.4 \mu\text{g m}^{-3}$.

Distribution of air mass types on the most polluted days

In winter, Polar Continental (AT 3) and AM-SCM (AT 4) occurred more frequently than expected on the most polluted days. These air masses are characterised by cold, anticyclonic conditions, with air originating from the north and the east. These conditions are conducive to temperature inversions and therefore pollutant stagnation. This supports the findings of McGregor (1999), which found that the sub-polar continental anticyclonic air mass type (similar to this study's Polar Continental) was disproportionately represented on the most polluted days. Conversely, Unstable Tropical Maritime (AT 2) is under-represented on days with high concentrations.

During the summer, Stable Anticyclonic Tropical (AT 3) and Tropical Continental (AT 5) disproportionately occur on the most polluted days for all pollutants. These air mass types are characterised by warm to hot temperatures and light wind conditions. A neutral to stable atmosphere suppresses vertical motion by impeding buoyancy and turbulent eddy motions. This leads to poor ventilation and reduced dispersion of pollutants (Cheng and Lam, 2000).

Severe pollution episodes

In winter, anticyclonic air masses Polar Continental (AT 3) and AM-SCM (AT 4) were found to be most favourable to severe winter pollution episodes. Crucially, McGregor et al. (1999) found that these two air mass types tend to occur together in a sequence during the winters of 1988-94 in Birmingham. Elevated respiratory response to pollutants can occur due to the changing nature of persistent winter anticyclonic weather from fine, cold and dry (AM-SCM) to overcast, cool and humid conditions (Polar Continental) advected from the north-east. Episodes of the AM-SCM air mass type creates a 'sensitising mechanism' by which cold, stable conditions occur in conjunction with elevated pollutant concentrations to prime the population's respiratory response, which is activated by the persistence of cool, moist flows of the Polar Continental air mass. While surface temperature inversions are a primary contributor to severe pollution days, inversions of nocturnal origin typically last only six to 12 hours; and are thus rarely a primary causal mechanism of winter severe pollution episodes (Kassomenos et al., 2007). High concentrations are unlikely to persist under the Tropical Maritime (AT 2) and Returning Polar Maritime (AT 6) air mass types, which proves that the advection of strong Atlantic westerlies disperses pollutants.

During the summer, Tropical Continental (AT 5) and Mild Stable Polar Continental (AT 4) were most susceptible to severe pollution episodes. These air mass types are often accompanied by an upper-level ridge and subsidence in the atmosphere. Severe summer O_3 pollution episodes only occurred under the Tropical Continental air mass (AT 5) type, a function of the high radiative input and calm conditions that characterise this airmass. Due to their cyclonic nature, Tropical Maritime (AT 1) and Transitional Polar Arctic (AT 2) produced few pollution episodes.

Cross-seasonal comparison of pollution episode severity

For all pollutants, the average duration of pollution episodes varies little between winter and summer. This similarity is likely due to distinctive air mass types in both seasons that are conducive to stability, namely Tropical Continental (AT 5) in summer, and Polar Continental (AT 3) and AM-SCM (AT 4) in winter. As expected, severe O_3 episodes last on average one day longer in summer because meteorological conditions favour increased photolysis, such as those experienced in the Tropical Continental air mass type.

Cross-seasonal comparison of pollution extreme values

With the exception of O_3 , all pollutants (and especially NO_x) have a significantly greater number of extreme pollution days in winter than in summer. This seasonal difference is explained by winter temperature inversions, which create enhanced atmospheric stability. In Birmingham, this occurs mainly through surface radiation inversions, in which the Earth's surface rapidly cools on clear and calm nights, creating a warmer boundary layer above the cooled surface. This suppresses atmospheric turbulence and mixing. It is significant that the two winter air mass types with the coolest temperatures and lowest wind speeds (AM-SCM AT 4 and Extreme Arctic Maritime AT 5) have on average the greatest pollutant concentrations (except O_3) of all 12 air mass types across winter and summer. This corroborates the findings of Widawski (2015), who showed that a marked temperature inversion in Warsaw (Poland) coincided with the highest levels of NO_x . Increased levels of NO_2 and CO in winter months can also be attributed to their significantly reduced influence on photochemical reactions occurring in photolysis for O_3 formation (Cichowicz et al., 2017). The meteorological contribution to significantly higher NO_x concentrations in winter may be of secondary importance. A three-fold increase in winter cloud cover would be expected to decrease photolysis such that the reduced reaction of NO_2 and sunlight should directly decrease NO_x production. Thus, the possible anthropogenic contributions to seasonal differences will now be explored.

Hamilton et al. (2009) indicate that the anthropogenic heat flux, the energy emitted from human activity in urban environments, almost doubles from summer to winter. During the winter, the use of electricity, central heating and gasoline (mainly vehicle engines) increases greatly because of the colder conditions. Gavin (2014) reported that electricity demand increased by 23% during the winter in the West Midlands. Moreover, fewer people walk and cycle in the winter and instead take to pollutant-producing vehicle transport (Yang et al., 2011). It follows that increased winter anthropogenic heat emissions in cities result in CO, PM10 and especially NO_x being released at a greater rate.

Pollutant concentrations found in this paper often exceed EC (2019) and WHO (2005) guidelines, especially in winter. This has significant implications for human health, given that those with respiratory diseases are at considerable risk during elevated concentrations of pollutants such as NO₂ and PM10. McGregor et al. (1999) demonstrated the relationship between respiratory-induced hospital admissions in Birmingham and air mass types. Their study revealed that PM10 concentrations exceeded the 50 µgm⁻³ threshold 22% of the time during anticyclonic polar continental air mass types, which correlated significantly with higher respiratory admissions. Pollutant concentrations also play a role in determining heart disease mortality rates in Birmingham (McGregor, 1999).

5. Conclusion

This study's primary aim was to examine whether a relationship exists between pollution and weather types, and whether this association varies seasonally. Principal component and cluster analyses were used to derive the air mass types, which were then related to pollutant concentrations. High concentrations are associated with anticyclonic types. In summer, such air mass types are characterised by low wind speeds, which leads to decreased advective mixing. During the winter, similar meteorological conditions are often combined with nocturnal radiative cooling. This creates a temperature inversion near the ground, which also suppresses the vertical dispersion of pollutants.

The tropical (polar) continental air mass type was most conducive to exceptionally high and very high concentrations in summer (winter). Turbulence is suppressed and the atmosphere is stable in these air mass types. In winter, a sequence of Polar Continental (cool and humid) and Binary Mid-latitude Anticyclonic Maritime – Sub-polar Cyclonic Maritime (cold and dry) induced severe pollution episodes in all pollutants, which can elevate the respiratory response in the general population by creating a sensitising mechanism.

The mean duration of a severe pollution episode duration varied little between winter and summer due to the strong presence of anticyclonic air mass types in both seasons. Naturally, increased photolysis results in severe O₃ episodes lasting 20% longer in summer than in winter. With the exception of O₃, high pollutant extremes were more common in winter. This was due to more favourable meteorological conditions and increased anthropogenic emissions during the cold season. Epidemiological evidence indicates that such high concentrations (e.g., one quarter of winter days exceeded the WHO and EC threshold for NO₂) can exasperate heart and respiratory diseases in Birmingham.

The main limitation of our paper is that it has analysed weather – pollution relationships at the daily timescale using the mean value of 24 hourly observations. In mobile situations, it is possible for multiple air mass types to occur on the same day. There was also an imperfection in the method used to calculate the wind component vectors U and V, which misrepresented some days' wind flow. For example, on the 23rd January, 2001, the wind direction was averaged as an easterly despite a five-hour 19mm rainfall event occurring under a strong westerly flow. Thus, the statistical analyses were unable to identify this rainfall event as being a westerly-induced phenomenon. However, a complete reversal in the wind direction (e.g., westerly to easterly; southerly to northerly) rarely occurs at the sub-daily timescale, which means that the above problem does not occur frequently. Some researchers (e.g., Cheng and Lam, 2000) overcame the problem of intra-diurnal variability by using six hourly averages. Whilst our paper uses 15 years of data, it is acknowledged that using linear regression to correct the values at Birmingham Centre to those at Tyburn between 2000-4 might have stripped out potentially extreme values at the latter site over this time period. This is because the regression line's position is constrained by having to pass through the bi-variate mean centre.

Future research should replicate the principal component and cluster analyses presented in this paper at the sub-daily timescale (e.g., six or 12-hourly means). Diurnal variations in pollutant concentrations would firstly have to be removed (e.g., calculate hourly standardised scores) before these sub-daily air mass types could be related to air quality data. Another avenue of further research could compare pollutant-air mass type relationships to health indicators, such as respiratory admissions or ischaemic heart disease data. Multi-variate analysis of variance could be used to test statistically the tripartite relationship between air mass types, pollution and concurrent health impacts rather than simply inferring it. Such research could be an invaluable device for authorities to forecast scenarios of concern to public health. This could be communicated to the public by means of a synoptic index. This paper has only used surface meteorological variables. Variables that capture vertical atmospheric stability (e.g., vertical temperature gradient) could be added to the analysis.

This study has clearly shown that an objective, synoptic air mass approach can be used to distinguish comprehensively the pollutant controls of different weather types. Given that pollutant guidelines were often breached during the study period, a concerted effort in the future to understand the complex relationship between meteorological conditions and anthropogenic-pollution relationship may have crucial epidemiological and societal implications.

Declarations

Ethics approval

The submitted work is original and has not been published elsewhere in any form or language.

Consent to participate

This paper is based on the analysis of secondary data sets. Hence, the only participants are the two authors.

Consent for publication

No persons were interviewed or questioned.

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Availability of data and material

Data sets used are freely available

Code availability

Not applicable

Authors' contributions

The first author obtained the data and performed all of the analyses. The second author edited the first author's work.

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Tables

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Figures

Figure 1

Case studies of the six winter air mass types Note: The designations employed and the presentation of the material on this map do not imply the expression of any opinion whatsoever on the part of Research Square concerning the legal status of any country, territory, city or area or of its authorities, or concerning the delimitation of its frontiers or boundaries. This map has been provided by the authors.

Figure 2

Case studies of the six summer air mass types Note: The designations employed and the presentation of the material on this map do not imply the expression of any opinion whatsoever on the part of Research Square concerning the legal status of any country, territory, city or area or of its authorities, or concerning the delimitation of its frontiers or boundaries. This map has been provided by the authors.

Figure 3

Air mass type pollution difference matrix for the winter. Pollutants that appear in the matrix show a statistically significant difference at the 99% confidence level between the pair of air mass types shown.

Figure 4

Air mass type pollution difference matrix for the summer. Pollutants that appear in the matrix show a statistically significant difference at the 99% confidence level between the pair of air mass types shown.

Figure 5
Comparison between the percentage occurrence of the air mass types across the whole winter and the percentage occurrence of days with high pollutant concentrations

Figure 5

Comparison between the percentage occurrence of the air mass types across the whole winter and the percentage occurrence of days with high pollutant concentrations

Figure 6
Comparison between the percentage occurrence of the air mass types across the whole summer and the percentage occurrence of days with high pollutant concentrations

Figure 6

Comparison between the percentage occurrence of the air mass types across the whole summer and the percentage occurrence of days with high pollutant concentrations

Figure 7
Number of severe winter pollution episodes per air mass and the relative dominance of that air mass in the episodes. The green line represents the percentage duration of pollution episodes in which the dominant air mass type (AT) is present.

Figure 8
Number of severe summer pollution episodes per air mass and the relative dominance of that air mass in the episodes. The green line represents the percentage duration of pollution episodes in which the dominant air mass type (AT) is present.

Figure 9

The arithmetic mean length in days of a severe pollution episode in summer and winter (bars), overlaid by their average concentrations (lines).

Figure 10: The number of days (y axis) that pollutant concentrations are exceptionally high, very high and high in summer and winter

Figure 10

The number of days (y axis) that pollutant concentrations are exceptionally high, very high and high in summer and winter

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