Small area spread and step-like changes in emergency medical admissions in response to an apparently new type of infectious event

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Abstract

Three potential outbreaks of an infectious-like agent are documented within 230 very small area geographies called mid and lower super output areas (MSOA, LSOA) in Wigan on the outskirts of greater Manchester in northern England from the early 1990’s onward, has allowed the identification of one such novel series of infectious-like events which are specific to a particular set of medical conditions apparently linked to infection, inflammation and autoimmunity [13-16]. Analysis of increased deaths and another four-in-a-row series occurred between 2008 and 2014 at two year intervals [17-19]. Increased deaths, medical admissions, emergency department attendances, NHS staff sickness absence and general practitioner referrals all show spatiotemporal spread during these infectious-like events [19-27], with full spatial spread across the UK appearing to take around two years while spread between the small areas which constitute a local authority take around 18 months [17-27].

This study analyzes very small area data covering the more recent 2008, 2010 and 2012 events for medical admissions to the Wigan Royal Infirmary located near the center of the Wigan borough council on the outskirts of greater Manchester in northern England. The bulk of these admissions are covered by a single primary care organization (PCO), the Wigan Borough Clinical Commissioning Group, hence any small area differences cannot arise from differences in acute admission thresholds, differences in PCO policies or practice, or to differences in adult social services. The findings from previous studies [24] are extended into very small area geographies called Lower Super Output Areas (LSOA) which contain around 100 medical admissions per...
annum, and use daily rather than monthly admissions to characterize a unique step-like increase in admissions which may best be described as a moving rectangular wave, in which admissions stay high for 12 months before rapidly reverting back to the baseline trajectory.

It has only recently been discovered that one of these events occurred in 2010 [18,19], and this event in particular can be investigated in more detail, along with the issue of spatiotemporal granularity/heterogeneity and infectious bursts which are a known feature of infectious outbreaks [28-31].

Methods

The data covers daily emergency (non-elective) medical admissions between April 2008 and March 2013, and comes from a previous study which used larger mid super output areas (MSOA), and examined the 2012 event in detail [24]. Medical admissions covers the specialties General & Elderly Medicine (55%), Emergency Assessment (18%), Thoracic medicine (11%), Cardiology (10%), Gastroenterology (6%) and Rehabilitation (1%). Admissions for this study were grouped into both the larger MSOA and smaller Lower Super Output Areas (LSOA), both of which are statistical geographies used in England. At that time the importance of the 2010 event was not appreciated [18,19], and hence this study also seeks to clarify the nature of this event.

No patient identifiable data was used, and a previously prepared extract at LSOA level [4] was analyzed. Admissions for persons living outside of Wigan were aggregated at local authority level or into larger geographies. Daily admissions for each MSOA/LSOA/area were summed into running 365 day totals. As in previous studies [19-25] the size of any step-like change is calculated by comparing the magnitude of the step-up and step-down associated with each event. This was determined using differences between successive running 365 day totals. Hence the first successive difference starts with a comparison of the running 365 day total ending 31st March 2009, with the running total ending 31st March 2010 as a percentage difference between the two periods. This detects any step-change occurring at 31st March 2009, move forward one day and repeat the comparison. Points of maximum and minimum percentage difference then identify the respective step-up and step-down events.

Monthly deaths (all-cause mortality) for the residents of Wigan were obtained from the Office for National Statistics. Data on the population density (persons/hectare), Index of Multiple Deprivation (IMD), which ranges from 0 (least deprived) through to 100 (most deprived), social classification (using the output area classification methodology), and proportion of persons aged 70+, were all obtained from the Office for National Statistics.

Statistical analysis is based on Poisson statistics where, by definition, the standard deviation is equal to the square root of the average. This square root relationship with the average attenuates the effect of uncertainty in the average. This study uses two measures of the average. In the evaluation of the maximum step-change over the entire period, the average was calculated over the full time period April-08 to May-12. For the evaluation of the 2009/10 event, the average was calculated over the interval specific to this event, namely, Mar-09 to Feb-11.

The relationship between the standard deviation of a Poisson distribution and the square root of the average can also be used to convert percentage increases in medical admissions in different sized LSOA (as average number of admissions) into standard deviation (STDEV) equivalents. Hence a 10% increase observed in a LSOA with an average of 100 admissions is equivalent to a 1 STDEV equivalent increase, etc.

Results

This study uses both MSOA and LSOA small areas within Wigan. Each MSOA contains three to eight LSOA with four to five being the most common. Each LSOA contains an average of 1,500 persons (range 1,000 to 3,000 persons, or 400 to 1,200 households). Hence this study is investigating very small spatial areas.

A running 365 day or 12 month total is a useful tool for detecting step-like changes in admission rates or deaths when the admissions/deaths show high seasonality [32-34], while the use of a larger running total minimizes the contribution from Poisson randomness, and effectively de-seasonalizes the trend [35]. A step-like increase, when analyzed using a running total, results in a ramp where the foot of the ramp marks the onset of the step-change, the slope of the ramp indicates the magnitude of the step-change, and the point 365 days or 12 months after the onset also reflects the size of the step-change (assuming the event has endured for a full 365 days).

Step-changes in deaths

To set the context for this paper, Figure 1 shows a running 12 month total of deaths for the 12 month period ending Dec-06 to May-15. According to the Office for National Statistics in the 2008-based components of change, deaths in Wigan were supposed to decline along a continuous line with around 3,100 deaths in 2009 reducing to 2,900 deaths in 2015 [36]. The peaks and troughs in Figure 1 (as seen in all other local authorities across the UK) are therefore entirely unexpected and have never been given an official explanation [15,18]. Figure 1 therefore shows a series of step-like increases in deaths commencing somewhere around May-07 (+8.7%), Feb-10 (+5.4%, less prominent perhaps due to slower and more granular spread), Apr-12 (+7.7%) and May-14 (+8.8%). Each peak may be made up from spatial spread within Wigan, and the remainder of this paper will investigate the accompanying small-area effect on medical admissions. It has been estimated that for every death there are around 10 extra medical admissions, although this ratio seems to vary between the events [37,38]. Note that background scatter of up to ±50 will arise due to Poisson variation and this explains most of the smaller saw-tooth features in Figure 1. The duration of this study does not cover the 2014/15 event although timing for initiation at Local Authority level across England has been presented elsewhere [19].

It has also been observed that deaths lag behind admissions and emergency department attendances by around one to two months [25,39], i.e. initial illness and decline precede eventual death. Hence the timing of small area events using medical admissions may occur earlier than seen for deaths in Figure 1.

Step changes in admissions

It is commonly believed throughout the health services in the UK, that population demographic change is the principle driving force for the increase in medical (and other) admissions [40], and an increasingly elderly population in Wigan would imply that the minimum number of admissions should occur toward the start of the study period and the maximum toward the end. This is totally at variance for the wide range in dates for the maximum and minimum admissions demonstrated in Table S1 in the supplementary material. Indeed the point of maximum 365 day total admissions can range anywhere between 31st Mar 2009 through to 31st March 2013 (Wigan 011C versus Wigan 032A), as can
the point of minimum admissions (Wigan 020A versus Wigan 012C). Typical winter infectious events are assumed to have only a small effect on the overall trend. The available data used in this study encompasses the tail end of the 2007/08 event, the full range of the 2010/11 event and the leading edge of the 2012/13 event, i.e. there are a maximum of three possible maxima and three possible minima during this period [19], as in Figure 1.

Admissions for residents in nearby West Lancashire displays three peaks of roughly similar magnitude for the 365 days ending 16th July 2009, 3rd March 2011 and 14th January 2013. For the residents of Wigan the tail of the 2008 event appears to merge with the start of the 2010 event giving an initial plateau around 17,650 admissions per 365 days, then a peak of 18,194 admissions for the 365 days ending 29th September 2010, followed by a minimum of 16,342 for the 365 days ending 31st January 2012 and then a maximum of 17,029 for the 365 days ending 12th November 2012 (data not shown).

However, as can be seen in Figure 2 the small-area trends display considerable variation between LSOA in Wigan. Such disparate trends are difficult to explain from an assumed demographic perspective, since all should be showing roughly linear trends with a slight upward slope of around 1% to 2% per annum. Hence as in Figure 1 there are a series of trends without any apparent explanation and which seem to defy commonly held notions regarding demographic growth as the principle driving force for change.

The initial plateau for Wigan as a whole, and Wigan 013C may illustrate the one weakness in the running 365 day approach in that an unexpected ‘spike’ event (arising from a typical temporary infectious outbreak such as influenza, SARS, or an extreme of temperature) creates a flat-topped plateau-like feature as the spike moves its way along the running 365 day total. The shape of the earlier part of the trend for Wigan 013C suggests such a spike event may be acting to obscure other trends, however, while the step-up may have been obscured, the step-down (as a ramp) is clearly visible. While it is far easier to adjust for spike events using monthly data (see [41] as an example), it is more difficult to do this in a running 365 day total, and it is exactly for this reason that this study uses both the step-up and step-down parts of the events, and for simplicity, this part of the analysis only documents the largest event in each small area. Hence using Figure 1 as an example the largest step-down would have occurred at the cessation of the 2007/08 event, etc.

Step-like increases in admissions (generating an upward ramp in a running 365 day total chart) can be seen, as are step-like reductions (a ramp downward) following the cessation of an event. The upward ramps are not perfectly linear simply because spread of the agent.

Figure 1. Running 12 month total of deaths (all-cause mortality) for residents of Wigan, near Manchester (UK) - this time-series covers four outbreaks of the suspected new agent.
throughout each LSOA is not instantaneous. Wigan 011C appears to show only the step-down following the 2007/08 event while Wigan 011B only appears to show a strong step-up for the 2012/13 event (the time span of the study limiting the ability to detect the ensuing step-down). As can be appreciated, the spatiotemporal trend for the whole of Wigan is a composite of all the smaller areas, and there appears to be considerable granularity/heterogeneity in the spread of the agent.

Table S1 in the Supplementary material presents summary data for the 100 largest LSOA. Note the large percentage difference between the maximum and minimum running annual total for admissions, which is far beyond anything arising from chance, and the wide variation in the dates for the maximum and minimum admissions, and the onset of the step-up and step-down features. Clearly something far more powerful than demographic-driven change is occurring and complex spatiotemporal effects are hidden in the larger geographic areas.

**Magnitude of the maximum step-up or step-down**

While the use of daily admissions generates more complex trends than those observed in earlier studies using monthly admissions, it is still possible to calculate the size of any step-like changes by comparing the percentage difference between successive 365 day blocks of data each side of a step-like change, for example, 365 days ending December 2012 compared to December 2011. Figure 3 shows the size of the largest step-like increase in each LSOA and MSOA (or other geography), and the size of the largest step-like decrease which will occur following the cessation of such an event. Both step-up and step-down have been calculated relative to the average number of admissions to avoid the situation where the step-down appears smaller by virtue of comparison against the larger total admissions after the start of the event. Figure 3 also shows the range between the maximum and minimum running 365 day total of admissions expressed as a percentage of the average admissions. As expected the range is slightly higher than the maximum step-up or step-down simply because it encompasses the sum of all possible effects against admissions during the time of the study. However the value of the range is not widely different to the step-up and step-down values, indicating that these step-like changes are the principle factor regulating changes in admissions.

As can be seen, the value for the largest step-up or step-down are roughly similar except at the extreme points measured across the entire hospital catchment area, or for the whole of Wigan. Also that the step-like changes are mostly greater than the value of a +1 standard deviation (SD) change which could arise from simple Poisson variation. As expected, potential Poisson variation increases with decreasing size since, by definition, the standard deviation of a Poisson distribution is equal to the square root of the average. In a Poisson distribution the +1 standard deviation line encompasses 85% of all possible outcomes arising from chance, and effectively represents the 85% confidence interval, i.e. in a Poisson distribution, 85% of all data values are expected to lie below the +1SD line – which is clearly not the case.

The value of the power function in the trend lines in Figure 3 range...
from -0.332 for step-down to -0.444 for step-up. It is possible that the value of the exponent has been reduced by the high data points at whole hospital and Wigan level. However these values are not too different from a value of -0.5 expected from a pure Poisson function, which indicates that Poisson chance events may form a basis for the mechanisms behind these events.

The value of ± 1 standard deviation (Poisson) can be used to determine the potential contribution from chance variation to each calculated increase. As can also be seen in Figure 3, the step-up in just four LSOA fails to exceed the 85% CI (Poisson), however, in all cases the maximum step-down is higher.

Table S2 in the supplementary material summarizes the details for the all the LSOA within the MSOAs in Wigan along with admissions from other areas. Note how the step-up or step-down is usually lower at MSOA level, and this is due to the wide variation in initiation dates for the step-up or step-down events, i.e. slow spatial spread within the geography of a single MSOA is leading to a degree of cancelling-out between the rectangular wave events which obscures the full extent of the effect seen at the smaller geographies. There are occasional exceptions to this rule and the maximum step-change in Wigan 022D is lower than that observed for Wigan 022. This is simply an expression of the high granularity/heterogeneity associated with these events.

**Timing of the maximum step-change**

The initiation dates for both the maximum step-up and step-down (12 months after initiation), can be used to demonstrate if the step-like events are clustered in time. From Figure 1 there may be the possibility of several step-down events after the 2007/08 event, the 2010/11 event has full coverage, while the 2012/13 event will be revealed by step-up features. Figure 4 summarizes the timing for the events captured at MSOA/LSOA level. As can be seen only six step-down events (with a maximum magnitude) were captured in the period Mar-08 to Aug-08, i.e. the bulk of the spatial spread associated with the 2007/08 event occurred too early for capture in this study.

However the 2010/11 event is well represented and appears to occur between Oct-08 through to Oct-10 or Mar-11 with greatest initiation occurring Aug-09 to Oct-09. The leading edge of the 2012/13 event probably overlaps the tailing edge of the 2010/11 event during the period Sep-10 to Mar-11, while the bulk of initiation commences after Jun-11 with a peak around Mar-12. Once again the duration of the study prevents full characterization beyond Mar-12.
The two year period of small area spread for the 2009/10 event revealed in this study, explains why the peak in deaths for this event in Figure 1 is not as sharp as the peaks seen for the other events in Figure 1. This concurs with other studies which suggest that the spread of deaths across the UK and Europe can occur more rapidly in some events than others [15,17,41]. It is clear that the mechanisms of spread, and any environmental factors leading to faster or slower spread, need to be understood.

Table S1 (supplementary material) shows the corresponding points in the running 365 day total at which the maximum or minimum total occurs (100 largest LSOA or other areas). As can be seen, the maximum or minimum can occur at any point in the four year period. Contrary to expectation the minimum occurs more frequently toward the end of the period than the start, i.e. population demography cannot be the major driving force for change. The most prominent minimum also occurs after the 2010 event but before the 2012 event.

**Effect of socio-demographic variables on the maximum step-change**

It is possible that socio-demographic variables may affect the magnitude of the observed step-change and roles for deprivation - as Index of Multiple Deprivation (IMD), population density, social grouping and age (as proportion of persons aged 70+) were all investigated. Figure 3 illustrates the importance of adjusting any step-change for the effects of size, and this was achieved by converting the magnitude of the step-change into standard deviation (STDEV) equivalents, i.e. magnitude of the step change divided by the square root of the average number of admissions.

Neither the IMD (range 4 to 73 in Wigan), or population density (range 2 to 90 persons per hectare in Wigan) showed any correlation with the magnitude of the step-change (see Table S3 in the supplementary material for a list of IMD, population density and social classification relating to each LSOA). However, as illustrated in Figure 5, social grouping appeared to show an effect with social groups containing older persons showing the greatest average step-change (as STDEV equivalents). Roughly similar differences in magnitude were observed between social groups in North East Essex following the 2008/09 event [21].

A potential role for age was further investigated by investigating the role of proportion of persons aged 70+ on the average maximum step-change, and is presented in Figure 6 where a running average of 10 size-ranked LSOA is used to minimize the scatter. The line of best

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**Figure 4.** Month of initiation of the events as deduced from maximum step-up and maximum step-down for these events in LSOA and MSOA in Wigan. Initiation derived from the step-down was assumed to occur 365 days earlier.
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The 2009/10 outbreak

In terms of the effect upon deaths across the UK, the 2010 event made only a small effect and would otherwise be assumed not to exist. It was only when this author was examining trends at Local Authority level that the possibility of an additional event began to emerge [18]. Given that the time span of this study allows complete characterization of the 2010 event, both the timing and magnitude of the step-increase was characterized in all MSOA/LSOA in Wigan. Since both step-up and step-down are available, a method was employed which looked for the point in time when the magnitude of the step-up and step-down reached a maximum. This method effectively determines the point at which the area under the inverted V arising from the running total method is maximized. The average value of the step-up and step-down associated with this point was calculated along with the associated date.

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Given the dependence on size noted in Figure 2, all percentage step-changes were adjusted to the theoretical value at an average of 100 admissions per annum, as follows: Observed step-change (%) x square root of (average admissions/100). The resulting size-adjusted step-changes ranged from +0% (Wigan 037C) to +198% (Wigan 017B). The average and median increases were 50% and 43% respectively while the interquartile range was 27% to 67%. Some 11% of LSOA experience greater than a 68% increase (2 x median) in medical admissions. Similar magnitude of increase in deaths has been characterized in equivalent sized locations in Australia [41].

As can be seen in Figure 7 the cumulative distribution is skewed. This skewing is evident in that at the left-hand tail 10% of locations experience a size-adjusted increase of between 0% and 18% (range 18%), while in the right-hand tail 10% of locations experience size-adjusted increases of between 86% and 198% (range 112%). In addition, the standard deviation is ± 29% which encompasses 9.1% of values in the left-hand tail, but only 0.4% of values in the right-hand tail. Hence this infectious-like event is clearly highly granular/heterogeneous with respect to its effect on increased medical admissions, and this behavior is consistent with the known behavior of infectious events [18].

As can be seen in Figure 8 there are three clusters of higher frequency of initiation around April/May 2009 (18% of locations), September/October 2009 (17%) and April/May 2010 (17%), while the frequency of initiation for other two month pairs only ranges from 8% to 12%. At the moment, the significance of the repeat April/May pairs toward the start and end of the outbreak will require further studies to

Figure 5. Effect of social group classification on the average maximum step-change converted into standard deviation equivalents.
see if this represents a significant period for these outbreaks in general. The median value of the (size-adjusted) step-increase in admissions also appears to peak at around the chronological mid-point of the outbreak, *i.e.* October and December 2009. A simple second order polynomial, using the ‘add trend line’ function in Excel, has been added to Figure 8 to show the trend in magnitude over time. This appears to be independent of summer/winter effects in that the median increase remains high from around July through to March. Given the known role of vitamin D in winter illness [43-48], it is worthy to note that vitamin D levels are at their highest at the end of summer [49], and due to the time lags in synthesis vitamin D only drops to a minimum toward the end of winter. Hence the effect seen in Figure 8 probably reflects the higher outbreak potential which would be expected in the mid-point of any outbreak.

### Discussion

This study builds upon three previous small area studies where the geographic areas are all within the larger area covered by a single acute site, by a single PCO and by a single social services [21-24,26-27]. The observed behavior therefore cannot be attributed to differences in acute thresholds to admission, PCO policies or practices, or to the funding of social services. Likewise differences in weather can be excluded - the geographies are simply far too close to each other to experience significant differences in weather-related variables.

The potential contribution of poor hospital care to the related increase in deaths which appears to accompany these events, can be categorically excluded on the basis that Hospital Standardized Mortality Rates (HSMR) at the Wigan Royal Infirmary were below average during the period of this study. As expected, there was a slight wobble in the HSMR during the one to two month period where deaths initially lag behind rising admissions.

A similar range in the dates for the maximum and minimum values of medical admissions seen in Table S1, has also been observed in Berkshire (south of England) for the same date range used in this study [22,26,27]. The key difference between these two locations appears to lie in the fact that the 2010 event seems to have had a far greater effect in Wigan than in Berkshire, while the reverse seems true for the 2012 event.

Most of the LSOA in Wigan are not subject to large population changes, and hence underlying growth is not large. The value of the calculated step-like changes shown in Figure 3 should therefore not require any growth adjustment.

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Figure 6. Effect of proportion of persons aged 70+ on the average step-change (STDEV equivalents) as running average of 10 proportion aged 70+-ranked LSOA.
A single clear step-increase would generate a ramp with little variation in slope over the ensuing 365 day period. From Figure 2 it is clear that the resulting ramp-like trends are more complex. From the perspective of disease transmission along social networks [29-31,50,51], an LSOA containing around 1,500 people is still a relatively large area, and given the implied relatively slow spread of the agent (i.e. relatively difficult to transmit) such complex time-trends are to be expected.

It must be recalled that a series of outbreaks at roughly two year intervals is an unprecedented situation since only two events per decade are most common [18]. Given that each event takes roughly two years to spread across the entire UK [18], it is therefore unsurprising that in some LSOA, late occurrences of the 2008 event can merge with the 2010 event and likewise with the 2010 and 2012 events. It has already been observed that spread occurs across the entire UK, and the member states of the European Union during an approximate two year window [15]. Similar dates and rate of spread have been observed in Australia [41] and New Zealand (in preparation). This wider spread precludes common health policy and funding since the four countries of the UK all operate different health policies and practices, as do the member countries of the European Union, while Australia and New Zealand are in the southern hemisphere.

The relationship between size and the apparent increase due to the step-like changes seen in Figure 3 has also been documented in Berkshire, and appears to arise from relatively slow spread of the agent within geographic areas containing different sized populations [22]. This is partly an expression of the Modifiable Areal Unit Problem (MAUP), and arises due to the slow spread within the different sized areas such that in larger areas there is greater opportunity for periods of maximum and minimum admissions within the constituent smaller areas to cancel each other out thereby reducing the value of the apparent step-increase [22]. This effect is illustrated in Table S2 (Supplementary material) where the step-increase is always greater for the LSOA within each MSOA than that for the MSOA.

A range of medical conditions are most affected, with a common linkage appearing to be related to immune function, i.e. infection, inflammation and autoimmunity [13,14,16,17]. In terms of increased death, the group of persons most dramatically affected by these events appears to be those suffering from existing neurological disorders, especially Alzheimer’s, dementia and Parkinson’s [52,53]. The very high prevalence of chronic fatigue, induced by peripheral inflammation and immune activation in this group, further emphasizes the potential pivotal role of immune dysfunction in susceptibility to these events [54].
Based on the range of conditions showing the greatest increase in admissions and deaths during these events, the ubiquitous immune-modifying persistent herpes virus, cytomegalovirus, has been implicated as either the causative agent or as secondary involvement by opportunistic reactivation in the presence of another agent [12-41,39,55-57]. Further work is required to validate this observation.

Factors influencing the increase in admissions

While population density has been shown to influence factors such as ambulance workload [58], the absence of a role for this factor in this study is not unexpected. This is because population density in Wigan is not high, and the maximum population density of 90 persons per hectare is only 14% of the maximum LSOA population density observed in London. At such low densities it is the social networks which will determine the degree of transmission [50,51]. Likewise deprivation using the IMD has been shown to influence health inequalities [59], however on this occasion it does not appear to be a major factor in transmission. From the evidence available, age appears to have the strongest factor influencing the percentage increase in admissions. Numerous studies have suggested that transmission of influenza-like illness depends mainly on contact with children who are the principal carriers [60,61], and it is contact with their elderly relatives which provides the opportunity for increased hospital admission among this age group. This risk of admission is then modified by the previous antibody history of the person to different strains of the agent [62].

Uncertainty in the initiation date

As has been discussed, two different methods were used to estimate the start of the 2009/10 event in each location. Hence either step-up or step-down or the method which assumes that step-down occurs 365 days after step-up and averages the value of the two events. In selecting the date for the maximum step-change it was noted that the maximum percentage change (to one decimal place accuracy) often occurred over a two to ten day range. For consistency the earliest date in this range was recorded as the start date. All methods detect the point of maximum 365 day change and therefore probably detect the point at which mass action has occurred, with actual arrival of the agent probably occurring days or weeks earlier. The resulting estimate of the (mass action) initiation date are listed in Table S2. Hence Wigan 001A has 17th October 2009 from all three estimates while Wigan 003A ranges from 1st Mar (from step-down) to 24th May (combined step-up/step-down). The assumption is that the event lasts for exactly 365 days, which may not always be the case. Table S2 also lists the time difference between step-up and step-down, and there is the possibility that a shorter duration than 365 days may be possible in some circumstances. This apparent duration could well be a function of the speed of transmission.
along the smaller social networks within each small area. Further work and mathematical modelling is required to clarify this issue.

**The particular shape of the 2009/10 event for deaths**

Given the detail now available from this study regarding the timing of the 2009/10 spatial spread throughout Wigan, the peculiar shape of the running 12 month total deaths in Figure 1 relating to the 2009/10 event can be explained. As has been explained a running 12 month total reveals the onset of a step-like increase in deaths or admissions observed as a ramp-up, and if this step endures for 12 months before abating, the running total then generates another ramp-down. This leads to an inverted ‘V’ shape in the running total. The base of the inverted ‘V’ should be 24 months wide. For the whole of Wigan the shape of the inverted ‘V’ can become corrupted if the underlying spatial spread of the agent is slow or patchy. The corrupted inverted ‘V’ in Figure 1 does however have a base which is 24 months wide from Jan-10 to Jan-12. In this study, the widest possible small area spread of the agent throughout Wigan has been shown to occur between Oct-08 to Apr-11 (Figure 4 – which used both step-up and step-down), while Figure 8 (which employed a method identifying greatest area under the inverted ‘V’ within a moving 24 month window) identified three periods of slightly higher spread over a more limited time period. Hence for whatever reason the 2009/10 event showed a high level of overlapping rectangular wave movement at small area level, the height of these wave being very different from one small area to another (Figure 7), all of which led to a corrupted inverted ‘V’ shape in Figure 1. It can only be assumed that similar slow and patchy spread across the whole of the UK lead to the 2009/10 event barely registering a 2% step-change, while the 2012 event shows a 5% step and the 2014 event shows a 9% step. The cluster of initiating events seen in Figure 4 around Mar-12 for the 2012 event suggests that particular events may spread faster than others, and this would lead to the range of inverted ‘V’ shaped running totals seen in Figure 1.

**Why have the events been overlooked?**

While it is not common to study medical admissions using very small area geographies, the considerable variation noted in this study would have resulted in great confusion to anyone attempting such analysis. Lacking knowledge of a potential infectious etiology, it is assumed that any researchers who attempted such analysis abandoned further study on the assumption that the apparently confusing results were simply due to statistical randomness or to errors in coding or data capture. The reliance on accurate patient information to trigger payment from the PCO to the hospital, excludes poor coding or data capture as an explanation for the trends seen in this study. Indeed neither coding nor data capture could account for the spatial spread between small areas.

**Effect on NHS Operational Efficiency and Finances**

These outbreaks pose considerable challenges in the area of health care finances and capacity planning at smaller local levels where there is considerable variation (see Figure 2 and Tables S1 and S2) [13,15,19]. However at the macro level observed by the national policy makers, the effect at larger geographies is considerably dampened, and this partly explains why these events have been overlooked for such a long period. This dampening effect is even greater using national rather than area data. Variations in the effect at larger geographies is considerably dampened, and this partly explains why these events have been overlooked for such a long period.

If these events are as significant as they appear to be from this study their effect should be felt across the entire NHS. In this respect spatial spread of increased hospital bed occupancy has been demonstrated for the 2002/03 and 2007/08 events [63], while a more recent study on bed occupancy for the whole of England demonstrated step-like increases in average occupancy for the 2009/10, 2012/13 and 2014/15 events [64]. At £300 per bed day it was estimated that the sudden increase in occupancy for the 2014/15 event led to a cost shock of around $121 million [64].

Indeed cycles in surplus and deficit have been documented in both the US health insurance industry and the NHS, where the start of the deficit part of the cycle appears to commence with the onset of these events [15,18,65-68]. A study using NHS emergency activity grouped into Healthcare Resource Groups (HRGs), the financial currency used in the NHS, showed a clear cycle in costs of around £600 million from trough to peak which was initiated by the 2002/03 and 2007/08 outbreaks [69]. The 2007/08 outbreak lead to a 6% to 36% increase in GP referrals to particular specialties observed at Health Board and PCO level across Scotland, Wales and England [20], which was age specific.

While there is now abundant evidence for the existence and impact of these events, sadly policy makers in the UK appear to wish to continue to blame the NHS for ‘inefficiency’ rather than acknowledge that a unique infectious event may be the real cause of the higher than expected long-term increase in medical admissions [15,70]. Indeed the large movements in admission implied by these events appear to be the fundamental source of year-to-year financial instability in the NHS [71-80].

**Future Research**

While this study has documented the small-area spatiotemporal behavior of medical admissions in some detail it raises a host of questions, such as:

1. What percentage of the increased admissions are related to persons in the last year of life?
2. Do the events increase the number of persons admitted and/or the number of readmissions per person?
3. What mechanisms underly the one year duration of the increased admissions?
4. To what extent is diagnostic ambiguity increased during these events?
5. Are there a set of biochemical or immune parameters which characterize those who are infected with the agent?

Hopefully further prospective and retrospective studies can answer some of these questions.

**Conclusion**

A series of events involving relatively slow spatial spread are clearly evident at small area level in Wigan and elsewhere in the UK. A powerful infectious-like agent is implicated which leads to large increases in medical admissions (and deaths) which endure for a period of around 12 months before abating. The size of the increase in medical admissions is more readily apparent in smaller population groups presumably because a large proportion of the population are infected at roughly the same time. Urgent research is required to identify the causative agent, and to model these outbreaks. Indeed any
agent capable of a sudden and sustained (size-adjusted) 50% increase in medical admissions needs to be given the utmost priority.

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