

1 **Landscape-scale exposure to multiazole-resistant *Aspergillus fumigatus***
2 **bioaerosols**

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20

21 **Abstract**

22

23 We demonstrate country-wide exposures to aerosolized spores of a human fungal pathogen,
24 *Aspergillus fumigatus*, that has acquired resistance to first line azole clinical antifungal drugs.
25 Assisted by a network of citizen scientists across the United Kingdom, we show that 1 in 20
26 viable aerosolized spores of this mold are resistant to the agricultural fungicide tebuconazole
27 and 1 in 140 spores are resistant to the four most used azoles for treating clinical aspergillosis
28 infections. Season and proximity to industrial composters were associated with growth of *A.*
29 *fumigatus* from air samples, but not with the presence of azole resistance, and hotspots were
30 not stable between sampling periods suggesting a high degree of atmospheric mixing. Genomic
31 analysis shows no distinction between those resistant genotypes found in the environment and
32 in patients, indicating that ~40% (58/150 sequenced genomes) of azole-resistant *A. fumigatus*
33 infections are acquired from environmental exposures. Due to the ubiquity of this measured

34 exposure, it is crucial that we determine source(s) of azole-resistant *A. fumigatus*, who is at
35 greatest risk of exposure and how to mitigate these exposures, in order to minimize treatment
36 failure in patients with aspergillosis.

37

38

39 **One sentence summary:**

40 UK-wide citizen science surveillance finds a ubiquitous exposure to aerosolized spores of a
41 human fungal pathogen that have evolved in the environment cross-resistance to essential
42 clinical antifungal drugs

43

44 **Manuscript**

45

46 The cosmopolitan mold, *Aspergillus fumigatus*, causes a spectrum of chronic and acute life-
47 threatening diseases in humans. The widespread occurrence of resistance to first line clinical
48 azole drugs in environmental isolates, alongside molecular epidemiology linking azole-
49 resistant *A. fumigatus* (ARAf) sourced from patients and the environment, argues that a
50 substantial burden of treatment failure is due to environmental azole exposure (1). The
51 numbers of patients in the United Kingdom (UK) presenting with infections that are resistant
52 to one or more of the clinical azoles is increasing in diverse patient groups (2, 3). The
53 significantly elevated case fatality rates where invasive aspergillosis is caused by ARAf (4, 5)
54 further highlights the importance and breadth of this emerging problem.

55

56 The widespread use of broad-spectrum agricultural fungicides, founded on the same
57 demethylase inhibitor (DMI) chemistry as the clinical azoles, has long been argued to drive the
58 evolution of environmental resistance (6). This hypothesis has found broad support from
59 surveillance demonstrating environmental hotspots of *A. fumigatus* growth alongside high
60 frequencies of resistance where this saprotrophic mold has the potential to grow in the presence
61 of agricultural DMIs (7). Yet, despite its increasingly wide detection in the environment
62 worldwide (7, 8), little is known about the extent to which humans are exposed to ARAf. The
63 mold is adapted to airborne dispersal and most humans inhale large numbers of viable spores
64 every day (9, 10). Owing to the potential clinical consequences of their inhalation, occupational
65 exposures to these spores are legislated in countries such as the UK, especially in green-waste
66 recycling and composting processes that have the potential to generate high levels of inocula
67 (11). However, occupational monitoring does not include assessing exposures to ARAf.

68 Surveillance has shown hotspots of environmental resistance in environments including both
69 home (8) and industrial compost (12), urban environments (13), greenhouses (14) and
70 horticultural products (15). Nonetheless, there is little insight into population-wide exposures
71 of at-risk individuals to aerosolised ARAf occurring beyond these heterogeneous
72 environmental foci.

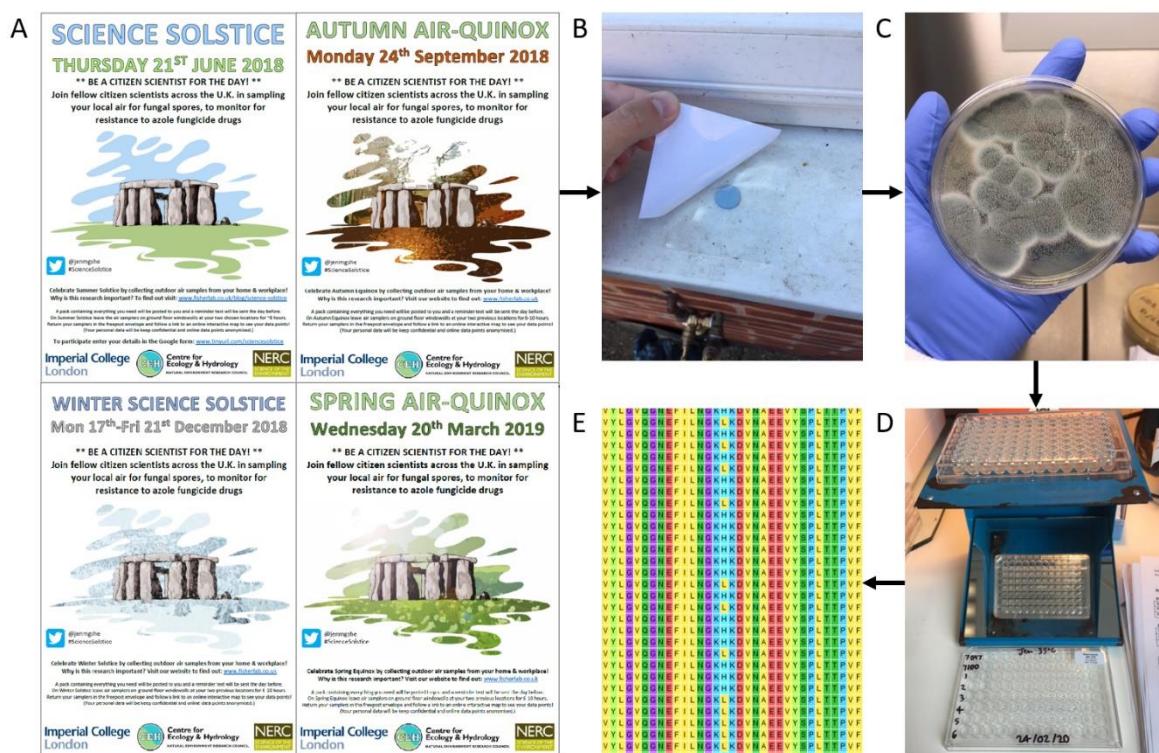


Figure 1: A) Posters advertising the four air sampling rounds that were posted on social media and displayed at Imperial College London and UK CEH, B) photo of a passive air sampler attached to an outdoor ground floor windowsill that had its sticky side exposed for 6-10 hours on a sampling day, C) isolates of *Aspergillus fumigatus* grown from an air sampler following incubation at 43°C for 48 hours, D) minimum inhibitory concentration (MIC) testing of tebuconazole-resistant *A. fumigatus* isolates to medical azoles, E) amino acid sequences spanning the L98H substitution for a subset of tebuconazole-resistant *A. fumigatus* isolates.

73
74 Given the dynamic nature of the atmosphere and the potential for season to affect the biology
75 of *A. fumigatus* spore production, meaningful assessment of human exposure to ARAf needs to
76 be undertaken at a population level in a cost-effective manner that spans yearly seasonal
77 variation. To meet this need, a UK-wide campaign was launched using social media, and used
78 to recruit a network of citizen scientists (16). These individuals then used simple passive air
79 samplers to collect airborne spores of *A. fumigatus* synchronously across a 6–10-hour time
80 period on the days matching the northern hemisphere seasonal equinoxes and solstices between
81 2018 and 2019 (**Figure 1A-C**). This activity resulted in a total of 1,894 air samples being
82 collected that, whilst being clustered owing to greater sample collection in areas of high

83 population density, achieved a near UK-wide distribution at each timepoint (**Figure 2A-D**).
84 These air samples were incubated on growth media using the highly selective temperature of
85 43°C and 919 (49%) yielded a combined 2,366 *Aspergillus* colonies. Of these, secondary
86 screening on media containing 6 mg/L of the commonly-used agricultural fungicide
87 tebuconazole (TEB; (17)) identified 111 TEB-resistant isolates, comprising 4.7 % of the total
88 isolates recovered (**Table 1**). This frequency is similar to that (~4%) measured at the
89 Rothamsted Research station in 2016 suggesting an incidence that is relatively stable across
90 recent years (18). Of the TEB-resistant isolates, 12 failed to sequence using the *cyp51A*
91 promoter and coding region primers and were re-identified by MALDI-TOF mass spectrometry
92 as the related species *Aspergillus lentulus* ($n = 10$) and *Aspergillus nidulans* ($n = 2$). For the
93 99 TEB-resistant isolates confirmed to be *A. fumigatus*, clinical breakpoints then showed that
94 85 (86%) were resistant to itraconazole (ITZ), the first-line clinical drug for chronic infections,
95 63 (64%) were resistant to voriconazole (VCZ), the first-line agent for invasive infections, 18
96 (18%) were resistant to posaconazole (PCZ) and 82 (83%) were resistant to isavuconazole
97 (ISZ) (**Figure 1D**). Of note, 50 (51%) of ARAf were resistant to three medical azoles and 14
98 (14%) were resistant to all tested medical azoles (**Table S1**), identifying a UK-wide aerosolized
99 exposure to drug-resistant variants of this pathogen.

100
101 In order to examine the genetic basis of the TEB-resistant phenotype, we genotyped the
102 canonical locus that confers azole-resistance, the sterol-demethylase gene *cyp51A* (19, 20).
103 Environmental azole resistance in *A. fumigatus* is most commonly due to within-gene point
104 mutations in *cyp51A* (**Figure 1E**) that are twinned with expression-upregulating tandem repeats
105 (TRs) in the promoter region (21). Genotyping confirmed that the most common aerosolized
106 resistance-associated polymorphisms were TR₃₄/L98H (59%) and TR₄₆/Y121F/T289A (6%).
107 We further found that 30% of TEB-resistant isolates did not contain any polymorphisms in the
108 *cyp51A* promoter or coding regions suggesting the existence of alternative resistance
109 mechanisms as have previously been noted (22) (**Table S1**). This is of concern in the clinical
110 setting as only the two former mutations are picked up by commercially available PCR
111 diagnostic methods, although previously unidentified mutations leading to resistance will be
112 picked up by phenotypic testing of minimum inhibitory concentration (MIC) providing that
113 there is an isolate available to test.

114

Sampling round	Number of air samples collected	Number of air samples that grew <i>A. fumigatus</i> (% of samples)	Number of <i>A. fumigatus</i> isolates grown from air samples	Average number of <i>A. fumigatus</i> isolates grown per air sample	Number of air samples that grew tebuconazole-resistant <i>A. fumigatus</i> (% of samples)	Number of tebuconazole-resistant <i>A. fumigatus</i> isolates (% of <i>A. fumigatus</i>)
summer	712	408 (57)	1,152	2.8	30 (4)	42 (4)
autumn	398	190 (48)	429	2.3	14 (4)	26 (6)
winter	320	152 (48)	477	3.1	12 (4)	15 (3)
spring	464	169 (36)	308	1.8	17 (4)	28 (9)
Total:	1,894	919 (49)	2,366	2.6	73 (4)	111 (5)

Table 1: The number of air samples collected, the number of air samples that grew *A. fumigatus* and the number of samples that grew azole-resistant *A. fumigatus* across the four air sampling rounds.

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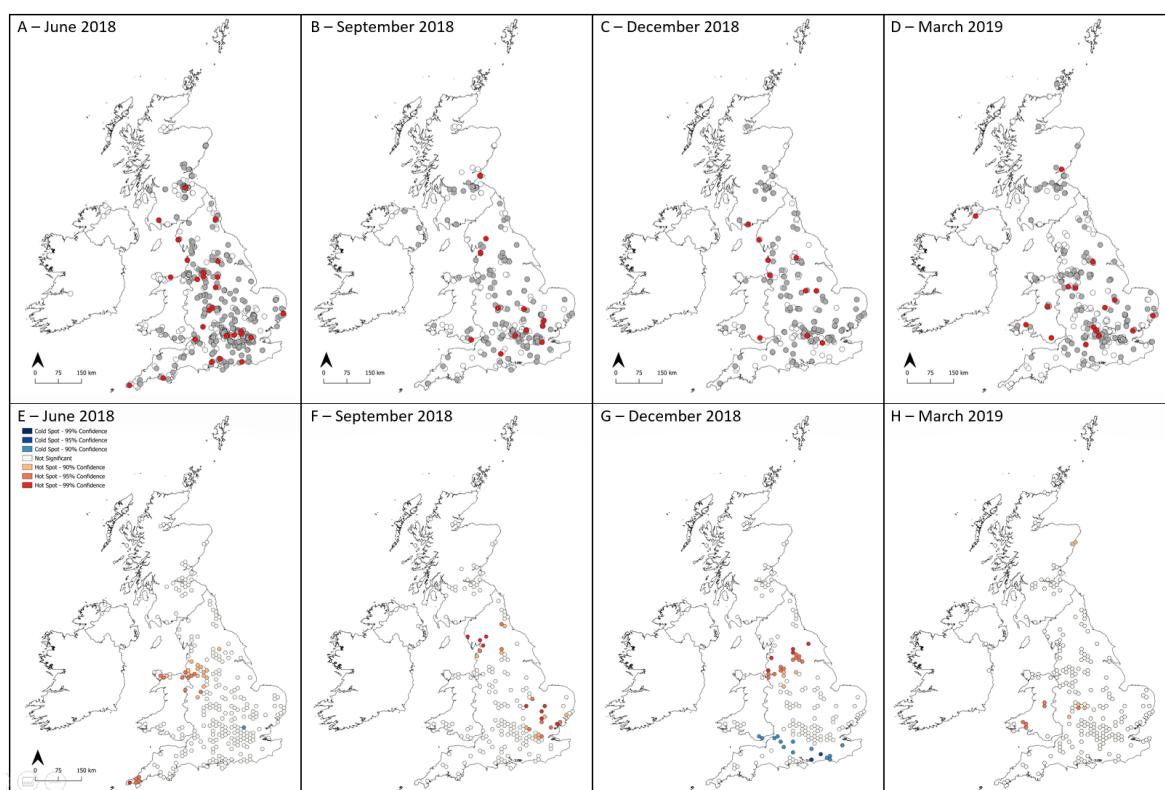


Figure 2. Maps showing locations of UK and Republic of Ireland air samples collected on A) 21st June equinox 2018, B) 24th September solstice 2018, C) 21st December equinox 2018 and D) 20th March solstice 2019. White dots indicate samples that did not grow *A. fumigatus*, grey dots indicate samples that grew tebuconazole-susceptible *A. fumigatus* and red dots indicate samples that grew tebuconazole-resistant *A. fumigatus*. Hotspots of azole-resistant *A. fumigatus* with 90%, 95% and 99% confidence according to Getis-Ord Gi* cluster detection analysis for sampling rounds on E) 21st June 2018, F) 24th September 2018, G) 21st December 2018 and H) 20th March 2019.

118

119 Subsequently, we determined the extent to which aerosolized ARAf matched those previously
120 recovered and sequenced from the UK terrestrial environments and patient cohorts by
121 sequencing the genomes of 62 ARAf isolates from the summer 2018 air sampling round (21
122 wild-type; 41 TEB-resistant). These data were then combined with those resulting from prior
123 UK-wide genomic surveillance (23) in a phylogenetic analysis. The resulting tree showed that
124 aerosolized isolates were broadly distributed throughout the UK phylogeny (**Figure 3A**) and
125 were drawn from both previously described clades 'A' (which contains the majority of ARAf
126 (23)) and 'B' (which is mainly sensitive to azole fungicides). Principal component analysis
127 (**Figure 3B**) corroborated our conclusion that there was no clear differentiation between
128 aerosolized isolates when compared to those from clinical and terrestrial sources, and
129 nucleotide diversity (π) tests showed that the observed genetic diversity separating these groups

130 was not significantly different (one-tailed t-test $p < 0.12591$). We did, however, note two
131 clusters of aerosolized isolates that were largely unrepresented in our previous surveillance,
132 suggesting a temporally dynamic aspect to the UK population genetic structure of this fungus
133 (**Figure 3A**). When mapped against the *A. fumigatus* reference genome *Af293*, each pair of
134 isolates across the combined dataset were, on average, separated by 24,000 SNPs, a figure that
135 was marginally lower (23,250 SNPs) when considering only aerosolized isolates. Significantly,
136 the genotype of an aerially sourced ARAf isolate grouped within a previously identified UK-
137 wide clonal subclade of genotypes, Clade A_A. Isolates in this clade all bear the hallmark
138 TR₃₄/L98H resistance allele and are widely found in the environment and infecting patients.
139 From these genomic data, we concluded that the genotypes of ARAf recovered from the UK
140 aerobiome are largely (but not exclusively) representative of those azole-resistant genotypes
141 recovered from patients. Moreover, these data indicate that ~40% (58/150 sequenced genomes)
142 of ARAf infections are acquired from environmental exposures.

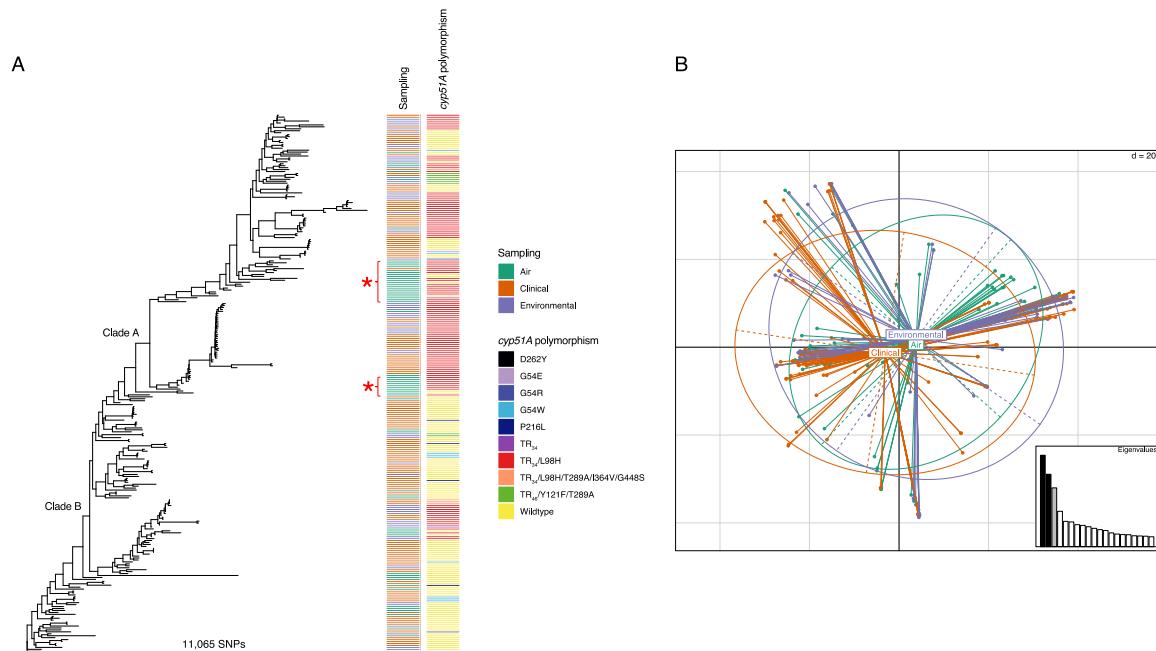


Figure 3. Phylogenetic and population genetic analyses indicate isolates sampled from air are not genetically distinct from the wider *Aspergillus fumigatus* population. A) Unrooted maximum likelihood phylogenetic tree (constructed in RAxML using genome-wide SNPs) showing the sampling type and *cyp51a* polymorphisms. Isolates sampled from air are found throughout the phylogeny. B) PCA indicate genetic identity of isolates, with air sampled isolates broadly deriving from the same wider *A. fumigatus* population.

143
144 We next sought to explain the spatial occurrence of *A. fumigatus* and ARAf using logistic
145 regression to determine which cardinal environmental variables (season, maximum daily
146 temperature, land cover classification and proximity to nearest industrial composting facility

147 with open windrow or outdoor activity (OW/OA)) affected their likelihood. Individually, each
148 variable was found to have a significant effect on whether a sample grew the mold (season χ^2
149 = 50.3, df = 3, $p < 0.01$; maximum daily temperature $\chi^2 = 29.3$, df = 1, $p < 0.01$; land cover
150 classification $\chi^2 = 17.4$, df = 7, $p = 0.01$; proximity to nearest OW/OA composting facility $\chi^2 =$
151 15.1, df = 1, $p < 0.01$; **Table 2**). Negative binomial regression then determined which
152 environmental variables affected the number of *A. fumigatus* colonies grown, with sampling
153 round ($\chi^2 = 35.9$, df = 3, $p < 0.01$) and proximity of sampling location to the nearest OW/OA
154 composting facility ($\chi^2 = 10.2$, df = 1, $p < 0.01$) showing significant associations. Subsequently,
155 Average Nearest Neighbour (ANN) tests found sampling locations in each sampling round to
156 be clustered (**Table 3**) and Getis-Ord Gi* spatial clustering analysis detected hotspots (high-)
157 and coldspots (low-prevalence) of airborne ARAf in each sampling round (**Figure 2E-H**).
158 However, none of the included environmental variables were found to affect the number of
159 ARAf grown from the air-samples, and the detected clusters were not stable between sampling
160 rounds.

161

Independent variable	Odds ratio (95% CI)	Pr(> z)
Sampling round		
<i>summer (baseline)</i>		
<i>autumn</i>	0.75 (0.57-0.99)	0.05
<i>winter</i>	0.94 (0.57-1.57)	0.82
<i>spring</i>	0.53 (0.37-0.75)	<0.01
Maximum daily temperature at sampling location on sampling date	1.04 (0.99-1.09)	0.13
Proximity of sampling location to nearest OW/OA composting facility	0.99 (0.99-0.99)	<0.01

Table 2: Odds ratios, confidence intervals and p-values calculated for independent variables included in a logistic regression model using air samples collected in the UK ($n = 1,894$) to explain whether a sample grew *A. fumigatus*. Significant results ($p \leq 0.05$) are highlighted in bold.

OW/OA = open windrow or outdoor activity.

162

163 In parallel to our aerosol sampling, a soil-sampling campaign was conducted by citizen
164 scientists during the 2019 summer solstice, resulting in the recovery of a high burden of ARAf
165 from garden soils totaling 736 (14%) resistant isolates from 246 locations (8). There were 46
166 participants from which both soil and at least one air sample were collected (**Supplementary**
167 **Table 2**). Of these, 23 (50%) grew ARAf colonies from either an air sample or soil sample, but
168 only 3 (7%) grew ARAf colonies from both. Moreover, there were no locations from which

169 ARAf was isolated from an air sample more than once. Taken together, these observations lead
170 us to conclude that the locally dynamic nature of atmospheric flows means that our bioaerosol
171 sampling strategy does not capture the presence of local ARAf soil hotspots. The corollary of
172 this observation is that atmospheric mixing leads to the UK population being, on average,
173 equally exposed to this bioaerosol. Based on our air sampling data and this assumption, we
174 estimate that the cumulative exposure of each individual across the UK to ARAf averages
175 22 (95% CI: 6-38) days per year.

176

	Sampling round				
	summer	autumn	winter	spring	all
Number of samples	712	398	320	463	1893
Observed Mean Distance (m):	3,059	3,505	3,314	2,945	855
Expected Mean Distance (m):	10,544	14,103	15,728	13,075	6,467
Nearest Neighbour Ratio:	0.29	0.25	0.21	0.23	0.13
z-score:	-36.24	-28.68	-27.01	-31.89	-72.23
p-value:	<0.00	<0.00	<0.00	<0.00	<0.00
Pattern	Clustered	Clustered	Clustered	Clustered	Clustered

Table 3: Results for Average Nearest Neighbour tests run for each sampling round separately and all sampling rounds together.

177

178 Importantly, while our study only reports on aerosolized spores collected in the UK, this
179 exposure is not restricted to the UK. Recruitment of citizen scientists attracted several
180 participants from outside the UK (24) with air samples from Germany, France and The
181 Netherlands growing viable ARAf. Fungal spores are readily dispersed across intercontinental
182 scales (25), and our population genetic studies of *A. fumigatus* has shown that its worldwide
183 distribution is unstructured, with no evidence for isolation-by-distance effects (26). Moreover,
184 the usage of DMIs in agriculture is widely increasing where measured (e.g., (27)), and
185 environmental azole-resistance has been detected in every country in which it was monitored
186 (28) indicating that ARAf has achieved a global occurrence. Yet, to date, there has been no
187 attempt to systematically measure population-wide exposures to ARAf more globally. This is
188 changing, and in 2022 the pan-South American LatAsp (<https://www.latasp.com>) surveillance
189 study commenced, mirroring the essential features of our Citizen Science campaign, with the
190 aim of determining the continent-wide incidence of ARAf. Nonetheless, identifying the
191 ecological hotspots that ultimately cause the miasma of azole-adapted mould that we document
192 will require spatially downscaled approaches if they are to be mitigated.

193

194 **Methods**

195

196 *Culturing Aspergillus fumigatus from UK air samples*

197

198 Air samples from which *A. fumigatus* isolates were cultured for this study were collected in
199 four sampling rounds of a citizen science project that took place between June 2018 and March
200 2019. The four sampling rounds took place on 21st June 2018 (summer solstice), 24th September
201 2018 (autumn equinox), 21st December 2018 (winter solstice) and 20th March 2019 (spring
202 equinox). In total, 485 individuals collected 1,894 air samples from England, Wales, Scotland
203 and Northern Ireland; 712 samples in summer, 398 samples in autumn, 320 samples in winter
204 and 464 samples in spring (**Table 1**).

205

206 Prior to the sampling date, citizen scientists were posted two passive air samplers (measuring
207 6.8 cm x 8.0 cm) to collect *A. fumigatus* spores, which were MicroAmpTM clear adhesive films
208 (ThermoFisher Scientific, UK) cut in half. The sticky side of each sampler was exposed
209 horizontally for 6-10 hours at approx. 1 m height, on the sampling date, re-covered and returned
210 by post to the primary author. When an air sample was received it was stored at room
211 temperature until processing, which involved removing the cover and placing the sampler
212 sticky-side down on a Sabouraud dextrose agar (SDA; Merck, Germany) plate containing
213 penicillin (Merck, Germany) at 200 mg/L and streptomycin (Merck, Germany) at 400 mg/L.
214 The plate was incubated at 43°C for 24 hours, the sampler was removed, and the plate was
215 incubated for a further 24 hours at 43°C. *A. fumigatus* isolates were picked one at a time using
216 a sterilized wooden toothpick into a tube containing mold preservation solution (MPS; 0.2%
217 agar and 0.05% Tween 20 in dH₂O) and stored at 4°C.

218

219 *Isolate screening for azole resistance*

220

221 Isolates were screened for tebuconazole resistance by pipetting 5 µl of MPS containing *A.*
222 *fumigatus* spores onto an SDA plate containing 6 mg/L tebuconazole, and a subset of isolates
223 (*n* = 250) were also tested using the Tebucheck protocol (17). The concentration of 6 mg/L
224 tebuconazole was chosen after testing the growth of 30 isolates with known *cyp51A* mutations
225 on SDA supplemented with 0 mg/L, 4 mg/L, 6 mg/L, 8 mg/L and 16 mg/L tebuconazole.
226 Isolates able to grow at a tebuconazole concentration of 6 mg/L were tested for susceptibility

227 to ITZ, VCZ, PCZ and ISZ according to CLSI M38-A2, as described in Borman *et al.* (2017).
228 Minimum inhibitory concentrations (MICs) were recorded as the lowest drug concentration at
229 which no growth was observed, and MICs were considered resistant when they were >1 mg/L
230 for ITZ, VCZ and ISZ and >0.25 mg/L for PCZ which are the suggested clinical breakpoints.
231

232 *Identification of A. fumigatus cyp51A gene azole-resistance alleles*

233

234 The promoter region of *cyp51A* was amplified using forward primer 5'-
235 GGAAGGGCTGATCAAACATATGC-3' and reverse primer 5'-
236 GTTCTGTTGGTTCCAAAGCC-3' and the PCR conditions: 95°C for five minutes; 30
237 cycles of 98°C for 20 seconds, 65°C for 30 seconds and 72°C for 30 seconds; 72°C for five
238 minutes. The PCR reaction volume used was 50 µl: 10 µl of FIREPol® DNA polymerase (Solis
239 Biodyne, Estonia), 10 µl of forward primer (1.5 µM; Invitrogen, US), 10 µl of reverse primer
240 (1.5 µM; Invitrogen, US), 18 µl of nuclease-free water (Merck, Germany) and 2 µl of DNA.
241 Amplicons were visualized by gel electrophoresis and samples with visible bands were sent for
242 sequencing using the forward primer. The coding region of *cyp51A* was amplified using
243 forward primer 5'-ATGGTGCCGATGCTATGG-3' and reverse primer 5'-
244 CTGTCTCACTTGGATGTG-3' and the PCR conditions: 94°C for two minutes; 35 cycles of
245 94°C for 30 seconds, 60°C for 45 seconds and 72°C for 45 seconds; 72°C for five minutes. The
246 PCR reaction volume used was 50 µl: 0.2 µl of Q5® high-fidelity DNA polymerase (New
247 England Biolabs, UK), 10 µl of Q5® reaction buffer (5X; New England Biolabs, UK), 0.5 µl
248 of deoxynucleotide (dNTP) solution mix (40 µM; New England Biolabs, UK), 1 µl of forward
249 primer (10 µM; Invitrogen, US), 1 µl of reverse primer (10 µM; Invitrogen, US), 35.3 µl of
250 nuclease-free water (Merck, Germany) and 2 µl of DNA. Amplicons were visualized by gel
251 electrophoresis and samples with visible bands were sent for sequencing in two segments using
252 the primers 5'-CTGATTGATGTCAACGTA-3' and 5'-
253 GATTCACCGAACTTCAAGGCTCG-3' (29). Sequences were aligned using Molecular
254 Evolutionary Genetics Analysis (MEGA) software (Penn State University, US).

255

256 *Identification of isolates*

257

258 Isolates that failed to sequence using the primers for the promoter and coding regions of *cyp51A*
259 were identified using matrix assisted laser desorption ionization-time of flight (MALDI-TOF)
260 mass spectrometry (MS), as described by Fraser and colleagues. (2016).

261 *Whole-genome sequencing of A. fumigatus isolates*

262

263 Genomic DNA (gDNA) was extracted from tebuconazole-resistant isolates with identity
264 confirmed as *A. fumigatus*. Isolates were revived from cryopreservation at -80°C by pipetting
265 20 µl into 25 cm³ NuncTM flasks containing SDA and incubating at 37°C for 48 hours. Spores
266 were harvested by washing the surface of the SDA with 10 ml of phosphate-buffered saline
267 (PBS) plus 0.01% Tween 20, 1.8 ml of spore suspension was added to 2 ml FastPrep tubes (MP
268 Biomedicals, US) and tubes were centrifuged at 5,000 rpm for 10 minutes. The supernatant
269 was discarded and the pellet resuspended in 300 µl of lysis solution and 1 µl of RNase A from
270 the MasterPureTM Complete DNA and RNA Purification Kit (Lucigen, US). The kit protocol
271 was followed, including an additional bead-beading step using a FastPrep-24TM instrument.
272 Extracted gDNA was purified using a DNeasy Blood and Tissue Kit (Qiagen, Germany) and
273 DNA concentration was measured using a Qubit fluorometer and Qubit dsDNA BR Assay kit
274 (ThermoFisher Scientific, UK). A NanoDropTM spectrophotometer (ThermoFisher Scientific,
275 UK) was used to assess DNA purity by checking that the ratio of absorbances at 260/230 nm
276 and 260/280 nm were 1.8-2.0. Purified gDNAs were stored at -20°C prior to being sent to
277 Earlham Institute (UK) where gDNA libraries were constructed, normalised and indexed.
278 Libraries were run on a NovaSeq 6000 SP v1.5 flow cell to generate 150 bp paired-end reads.
279 These data are deposited in the European Nucleotide Archive (ENA) under Project Accession
280 PRJEB51237.

281

282 All raw reads were quality checked with FastQC v0.11.5 (Babraham Institute) and aligned to
283 the reference genome Af293 (30) using Burrows-Wheeler Aligner v0.7.8 MEM (31) before
284 conversion to sorted BAM format using SAMtools v1.3.1 (32). Variant calling was performed
285 with GATK HaplotypeCaller v4.2.6.1 (33), excluding repetitive regions (identified by
286 RepeatMasker v4.0.6), generating GVCFs. Low-confidence variants were filtered providing
287 they met at least 1 of the parameters DP < 5, GQ < 50, MQ < 40, MQRankSum < -12.5,
288 ReadPosRankSum < -8.0, SOR > 4.0. In addition, alternate variants must be present in at least
289 90% of the reads. SNPs were mapped to genes using vcf-annotator (Broad Institute).

290

291 Phylogenetic analysis was carried out on 62 tebuconazole-resistant *A. fumigatus* isolates
292 collected by air sampling in addition to 215 environmental and clinical *A. fumigatus* isolates
293 with complete sampling history collected in the UK between 2005 and 2017 (23) (data
294 available from PRJEB27135 and PRJEB8623). Whole genome SNP data were converted to

295 presence/absence of a SNP with respect to reference, and any SNPs identified as low
296 confidence in the variant filtration step were assigned as missing. These data were converted
297 to FASTA format. Maximum-likelihood phylogenies were constructed using rapid bootstrap
298 analysis over 1000 replicates using the GTRCAT model of rate heterogeneity in RAxML v8.2.9
299 (34) to assess sequence similarity between isolates, and resulting phylogenies were visualized
300 using ggtree v3.1.14 and Microreact
301 (<https://microreact.org/project/6NMrDobYGZnhmYnMSsHhC5-air>) (Figure S1).

302
303 Genetic similarities were investigated using the hypothesis-free approaches Principal
304 Component Analysis (PCA) and Discriminant Analysis of Principal Components (DAPC) (35)
305 using the R package adegenet v2.1.5 in R v4.1.0. Nucleotide diversity tests were implemented
306 in VCFtools v0.1.13 (36).

307
308 *Environmental variables thought to influence growth of A. fumigatus*
309

Environmental variables ascertained for sampling date and location	Source of information
Maximum daily temperature at sampling location on sampling date (°C)	Met Office HadUK-Grid dataset (37)
Land cover classification of sampling location (21 categories)	UKCEH Land Cover Map 2019 (38)
Distance of sampling location to nearest composter with open windrow or outdoor activity (OW/OA) (m)	Composter locations obtained from Environment Agency, Scottish Environment Agency (SEPA) website (39), Natural Resources Wales website (40) and Northern Ireland Environment Agency website (41). Distances calculated using package “geosphere” in R version 4.0.0.

310
311 **Table 4:** Environmental variables obtained for air sampling locations and dates and the sources they were
312 obtained from.

313
314 **Table 4** details the environmental variables that were ascertained for the sampling dates and
315 locations, based on the information provided by citizen scientists. Sampling round was used as

313 a proxy for season: 21st June 2018 (summer), 24th September 2018 (autumn), 21st December
314 2018 (winter), 20th March 2019 (spring).

315

316 *Generalized linear models*

317

318 Generalized linear models (GLMs) run in R v4.0.0 (42) were used to find associations between
319 the environmental variables in **Table 4** and i) the likelihood of a sample growing *A. fumigatus*,
320 and ii) likelihood of a sample growing ARAf. Growth of *A. fumigatus* or ARAf from a sample
321 was categorized as 0/1 and logistic regressions (“glm” function; family = “binomial”) were
322 performed. Significant improvement on the null model, as determined by analysis of variance
323 (ANOVA) using chi-squared test, determined which environmental variables were included in
324 the regression model. Reduced Akaike information criterion (AIC) score and significant
325 improvement on the null model were used to choose the regression model with the best fit.
326 Results were considered significant when $p \leq 0.05$.

327

328 *Spatial clustering analysis*

329

330 ArcMap 10.7 was used to investigate the spatial distribution of all sampling locations and the
331 samples that grew ARAf. Sample locations were georeferenced and projected using the British
332 National Grid (EPSG:27700) coordinate reference system.

333

334 The Average Nearest Neighbour (ANN) test was used to determine whether the sampling
335 locations in each sampling round, and across all sampling rounds, were randomly distributed
336 or spatially autocorrelated. This test measures the distance between each sampling location and
337 the next closest sampling location and compares this to the distance that would be expected if
338 the locations were randomly distributed throughout the study area (the UK). If the observed
339 mean distance is less than the expected mean distance, the spatial distribution of the observed
340 data is considered spatially clustered. Alternatively, if the mean distance is greater than the
341 expected distribution, the features are considered dispersed. In both scenarios, ANN ratios were
342 used to establish the type of spatial pattern (*i.e.* clustered, dispersed or random) and associated
343 p values <0.05 were used to establish statistical significance.

344

345 A local indicator of spatial autocorrelation was used to detect and geographically identify
346 hotspots of azole resistance in the UK. First, sampling locations were geospatially aggregated

347 within a gridded hexagon structure (cell area 115km²). For each cell, the total number of
348 collected samples and the total number of azole-resistant *A. fumigatus* samples were extracted
349 and used to calculate sample positivity (%). Cells with zero sample locations were removed
350 prior to analysis. Then, a Getis Ord Gi* analysis was conducted to identify statistically
351 significant hotspot cells based on sample positivity. This approach was used to generate local
352 z scores for each cell, with statistically significant high and low values indicating hotspots and
353 coldspots, respectively. For this analysis hot- and coldspot cells were generated and grouped
354 using three different significance values (90, 95 and 99%).

355

356 *Average exposure simulation analysis*

357

358 Assuming that any location could potentially be ARAf positive at some point in time, we
359 conducted a simulation based on 1,000,000 individuals, each having a respiratory rate extracted
360 from a normal distribution with a mean of 9L/min and a standard deviation of 2L/min. To
361 match the air sampling campaigns conducted, we considered 1,095 time periods of 8 hours
362 over a period of one year (365 days). The probability of each individual being exposed to ARAf
363 was based on a truncated normal distribution of mean 5.86% and standard deviation of 2.80%,
364 set with 0 as the lower bound. We then calculated the number of 8-hour ARAf positive period
365 for each individual, from which we derived the annual average across our total population.

366

367 **Acknowledgements**

368 The authors would like to thank all the citizen scientists who collected air samples for this
369 study. We also thank Dr Pippa Douglas for providing the locations of composters in England
370 with open windrow or outdoor activity, and Dr Jianhua Zhang for sharing the *cyp51A* coding
371 region primers.

372

373 **Funding Information**

374 This work was supported by the Natural Environment Research Council (NERC;
375 NE/L002515/1 and NE/P000916/1) and the UK Medical Research Council (MRC;
376 MR/R015600/1). MCF is a fellow in the CIFAR ‘Fungal Kingdoms’ program. AA was
377 supported by a postgraduate studentship from Al-Baha University, Saudi Arabia.

378

379 **Competing Interests**

380 The authors have no competing interests to declare.

381

382 **Author Contributions**

383 JMGS, ACS, and MCF conceptualised the study; AA, PSD, MF, AMB and EMJ contributed
384 experimental techniques; JMGS, SH, APB and TRS processed samples; JMGS, JR, CBU and
385 FBP analysed the data; JMGS, JR and MCF drafted the original manuscript, which ACS, EMJ,
386 SH and FBP reviewed and edited.

387

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389

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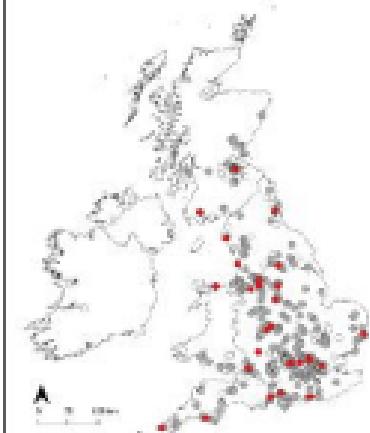
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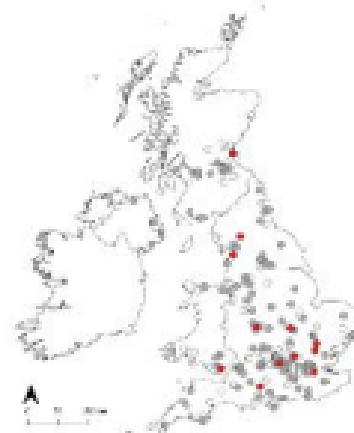
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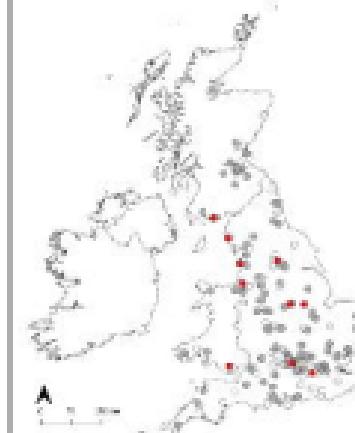
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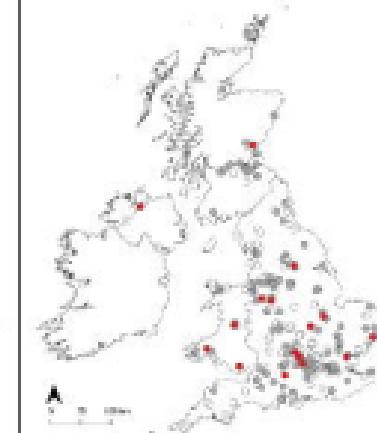
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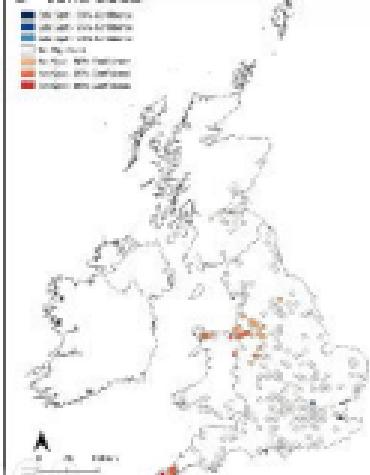
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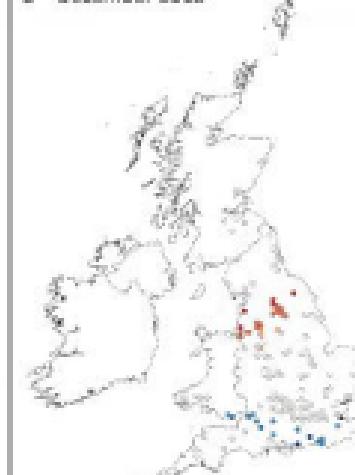
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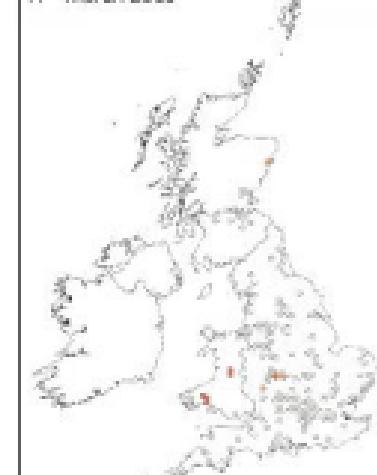
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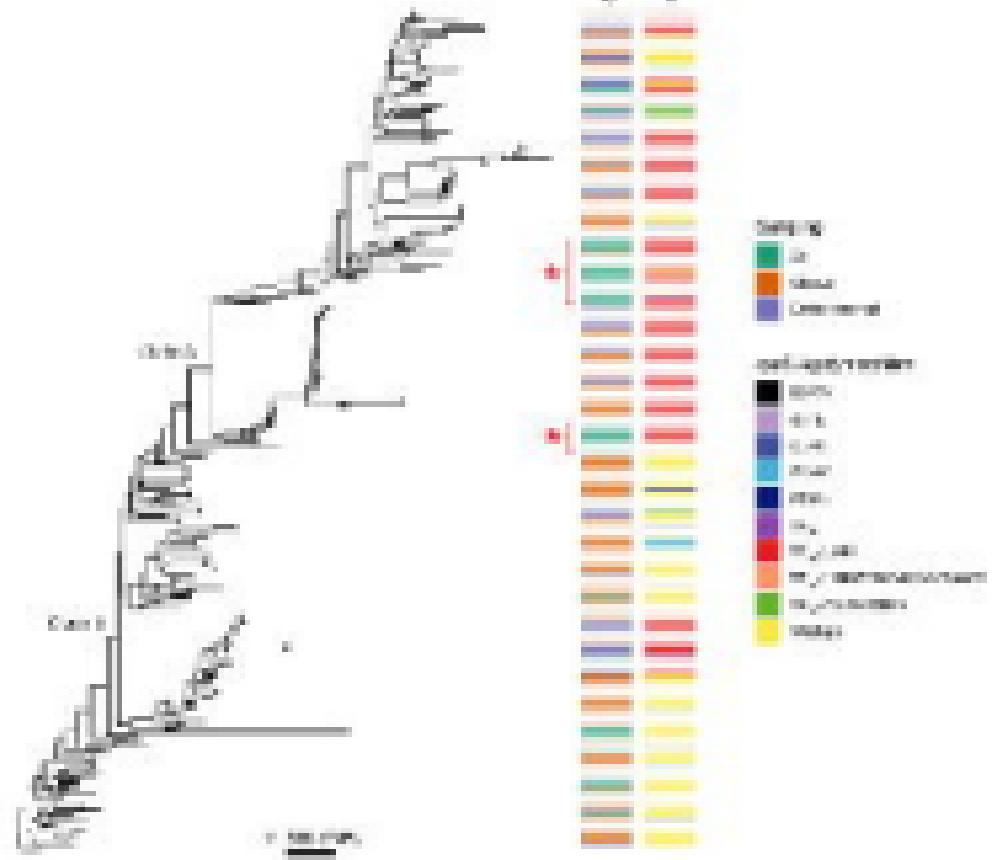
G – December 2018



H – March 2019



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