Genome-wide changes in genetic diversity in a population of

Myotis lucifugus affected by white-nose syndrome

Thomas M. Lilley,[*,†,1] Ian W. Wilson,[*] Kenneth A. Field,[‡] DeeAnn M. Reeder,[‡] Megan E.

Vodzak,[‡] Gregory G. Turner,[§] Allen Kurta,[**] Anna S. Blomberg,[††] Samantha Hoff,[‡‡] Carl J.

Herzog,[‡‡] Brent J. Sewall,[§§] Steve Paterson[*]

[*]Institute of Integrative Biology, University of Liverpool, United Kingdom. [†]Finnish Museum

of Natural History, University of Helsinki, Finland. [‡]Biology Department, Bucknell

University, 1 Dent Drive, Lewisburg, PA 12837, USA. [§]Pennsylvania Game Commission,

Harrisburg, PA, USA. [**]Department of Biology, Eastern Michigan University, Ypsilanti, MI,

USA. [††]Department of Biology, University of Turku, Finland. [‡‡]Wildlife Diversity Unit, State

Department of Environmental Conservation, Albany, NY, USA. [§§]Department of Biology,

Temple University, Philadelphia, PA, US

Raw data are available at ENA accession ERP120680, and bioinformatics code at

https://github.com/scottishwormboy/myoluc.

Running title: Effects of fungal pathogen on genomes of bat host population.

Keywords: Opportunistic pathogen, white-nose syndrome, Selective pressure, Genetic

resistance.

[1]Corresponding author: Thomas M. Lilley, Finnish Museum of Natural History, University of

Helsinki, P. Rautatiekatu 13, PL17, 00014 Helsinki, Finland.

Phone: +358408337783

Email: thomas.lilley@helsinki.fi

ABSTRACT

Novel pathogens can cause massive declines in populations, and even extirpation of hosts.

But disease can also act as a selective pressure on survivors, driving the evolution of

resistance or tolerance. Bat white-nose syndrome (WNS) is a rapidly spreading wildlife

disease in North America. The fungus causing the disease invades skin tissues of hibernating

bats, resulting in disruption of hibernation behavior, premature energy depletion, and

subsequent death. We used whole-genome sequencing to investigate changes in allele

frequencies within a population of *Myotis lucifugus* in eastern North America to search for

genetic resistance to WNS. Our results show low F_{ST} values within the population across

time, i.e. prior to WNS (Pre-WNS) compared to the population that has survived WNS (Post-

WNS). However, when dividing the population with a geographical cut-off between the

states of Pennsylvania and New York, a sharp increase in values on scaffold GL429776 is

evident in the Post-WNS samples. Genes present in the diverged area are associated with

thermoregulation and promotion of brown fat production. Thus, although WNS may not have

subjected the entire *M. lucifugus* population to selective pressure, it may have selected for

specific alleles in Pennsylvania through decreased gene flow within the population. However,

the persistence of remnant sub-populations in the aftermath of WNS is likely due to multiple

factors in bat life history.

45   INTRODUCTION

The emergence and spread of multiple infectious wildlife diseases during recent decades has

had devastating consequences for biodiversity (Daszak *et al.* 2000; Smith *et al.* 2006).

Unfortunately, anthropogenic threats, including human-mediated introductions and climate

change, appear to be the main causes of exposure to sources of infection (Gallana *et al.* 2013;

Martel *et al.* 2014; Tompkins *et al.* 2015). With shifting pathogen distributions, disease-

related declines in naïve wildlife often threaten the persistence of populations. Examples

include chytridiomycosis, which decimated populations of amphibians globally (Daszak *et al.*

1999; Lips *et al.* 2006), and avian malaria, which caused the sharp decline of island

populations of birds (van Riper *et al.* 1986). More recently, white-nose syndrome (WNS) has

been described as one of the most rapidly spreading wildlife diseases ever recorded (Blehert

et al. 2009; Frick *et al.* 2010). Since the discovery of WNS in North America in early 2006,

13 species of bats have been diagnosed with the disease in 34 U.S. states and 7 Canadian

provinces (www.whitenosesyndrome.org 2020). Genetic evidence suggests that

Pseudogymnoascus destructans, the causative agent of WNS, was introduced to North

America from Europe (Wibbelt *et al.* 2010; Ren *et al.* 2012; Lorch *et al.* 2013; Minnis and

Lindner 2013; Campana *et al.* 2017), where affected species do not experience associated

mortality (Puechmaille *et al.* 2011; Warnecke *et al.* 2012; Wibbelt *et al.* 2013; Zukal *et al.*

2016; Harazim *et al.* 2018).

When a disease enters a naïve host population, the initial wave of infection often causes

epizootics resulting in mass mortality, which may extirpate local host populations or even

cause species extinction (Daszak *et al.* 1999; De Castro and Bolker 2005; Frick *et al.* 2010).

Where host extirpation does not occur, disease may instead act as a selective pressure on

survivors, driving the evolution of tolerance or resistance and transforming a disease from

being epizootic to being enzootic (Boots *et al.* 2009; Robinson *et al.* 2012; Karlsson *et al.*

2014). Where selective pressure is strong, this may occur through rapid changes in the

distribution of genetic variants associated with disease susceptibility over short timescales

(Gallana *et al.* 2013) and may be detectable for generations (Groot *et al.* 2002; Di Gaspero *et*

al. 2012; Sironi *et al.* 2015; Deschamps *et al.* 2016). The detection of selective sweeps on

particular genes or gene families has been proposed to confirm or exclude potential

mechanisms of host susceptibility or pathogen virulence (Campbell and Tishkoff 2008).

However, rapid bottlenecks (such as those caused by panzootic events) are associated with a

more stochastic loss of alleles, which does not necessarily indicate selection (Luikart *et al.*

1998).

The psychrophilic (cold-growing) fungus *P. destructans* (Minnis and Lindner 2013) that

causes WNS (Lorch et al. 2011) acts as an opportunistic pathogen of bats, invading the skin

tissues of hibernating hosts (Cryan *et al.* 2010; Meteyer *et al.* 2012). Susceptible species,

such as *Myotis lucifugus* and *M. septentrionalis*, have shown population declines greater than

90% in affected hibernacula (Frick *et al.* 2015). The infection disrupts hibernation behavior

of susceptible species and leads to more frequent arousals from torpor, evaporative water

loss, premature energy depletion, and death of susceptible individuals due to emaciation

(Willis *et al.* 2011; Reeder *et al.* 2012; Warnecke *et al.* 2013; Verant *et al.* 2014; McGuire *et*

al. 2017). Naïve infected *M. lucifugus* upregulate genes involved in immune pathways during

the hibernation period (Field *et al.* 2015, 2018; Lilley *et al.* 2017). These responses are weak

during torpor but are robust during the intermittent arousals (Luis and Hudson 2006; Field *et*

al. 2018). Therefore, increased arousals may be attempts by the host to counter the pathogen,

in addition to quenching thirst, grooming, expelling waste and possibly foraging (Willis *et al.*

2011; Brownlee-Bouboulis and Reeder 2013; Bernard and McCracken 2017), and

supplementing electrolytes (Cryan *et al.* 2013).

Much of the research on disease-induced selection has focused on the major

histocompatibility complex (MHC), and indeed, diseases can drive the evolution and

maintenance of MHC diversity in natural host populations (Paterson *et al.* 1998; Jeffery and

Bangham 2000; Teacher *et al.* 2009; Spurgin and Richardson 2010; Zeisset and Beebee 2014;

Davy *et al.* 2017). Yet, factors not directly associated with interactions between host and

pathogen, such as environmental conditions and competition with other species, can have a

considerable influence on the manifestation of a disease (Scholthof 2007). In particular,

white-nose syndrome is a prime example of a disease that is manifested when factors

associated with the environment (i.e. temperature and humidity of the hibernaculum), the host

(and the host's response to infection) and the pathogen (optimum growth temperature range,

suitable host) overlap, i.e. intersect within the "disease triangle" (Scholthof 2007; Turner *et*

al. 2011). Therefore bat species, and populations within species, are affected differently

according to hibernation behavior and prevailing micro-climate conditions in available

hibernacula (Johnson *et al.* 2014; Langwig *et al.* 2015; Grieneisen *et al.* 2015). As such, in

hosts challenged with an opportunistic pathogen, such as *P. destuctans*, that is capable of

persistence in the environment in the absence of the host, loci not associated with MHC

diversity or other immune response-associated factors (Donaldson *et al.* 2017), such as the

amount of body fat prior to hibernation (Cheng *et al.* 2019), may also affect survival of

species and populations. This highlights the importance of examining the genome as a whole

(Sparks *et al.* 2019).

The initial impacts of WNS on six species of hibernating bats in the northeastern and

midwestern USA have varied from almost complete extirpation to arrested population growth

at the colony scale (Turner *et al.* 2011), leading to extensive declines at larger geographic

scales (Thogmartin *et al.* 2012; Ingersoll *et al.* 2013). After the initial decline caused by

WNS in *M. lucifugus*, reports have surfaced describing stabilization of colonies at smaller

sizes or even increases in numbers of individuals in some areas (Langwig *et al.* 2015; Maslo

et al. 2015; Frick *et al.* 2017; Dobony and Johnson 2018). Models parameterised with long-

term data on fungal loads, infection intensity and counts of *M. lucifugus* at hibernacula have

suggested development of either tolerance or resistance in these persisting populations (Frick

et al. 2017). This is supported by reports of infected individuals not arousing from torpor as

frequently as during the acute phase of the zoonosis (Lilley *et al.* 2016; Frank *et al.* 2019).

Because WNS has resulted in massive population declines in some *M. lucifugus*, there is a

possibility that selection could occur for alleles conferring resistance or tolerance within the

standing genetic variation. Indeed, Maslo and Feffermann (2015) suggested the occurrence of

evolutionary rescue, and Donaldson et al. (2017) described changes in an immunome across

M. lucifugus populations obtained through a sequence capture method, which may be

attributed to selection by WNS. Langwig et al. (2017) suggested that the initially affected *M.*

lucifugus population is beginning to show signs of resistance to the pathogen. However, the

selective pressures WNS has exerted on the population, which may not even be related to

immune responses (e.g. Cheng et al., 2019), have only recently been described at the whole-

genome level (Auteri and Knowles 2020), and high-sequencing depth studies are still lacking.

Furthermore, no knowledge exists on how WNS may have affected gene flow within the

previously panmictic eastern population of *M. lucifugus* (Miller-Butterworth *et al.* 2014;

Wilder *et al.* 2015) and if responses differ between bats in different geographic areas.

Here, we utilize high-throughput whole-genome sequencing of the WNS-affected species *M.*

lucifugus, allowing us to look at entire genomes via single nucleotide polymorphisms (SNPs).

We compare genetic patterns (F_{ST} and heterozygosity) before and after the spread of WNS in

the eastern North American population to gain insight into whether the disease is causing

selection of major loci in surviving bats. We also examine whether the disease may have

decreased gene flow within the previously panmictic population, and if bats in two

geographic areas show differing signs of selection, by comparing samples collected from

Pennsylvania to samples collected from New York. We hypothesize that, due to the massive

population declines in *M. lucifugus* caused by WNS at a large geographic scale (Ingersoll *et*

al. 2016), and the possibility that the affected population may now be beginning to stabilize

or even slightly increase in size in some areas (Langwig *et al.* 2017), selection based on

standing genetic variation has resulted in differentiation in one or many regions of the

genome (Messer and Petrov 2013). We also hypothesize that reduced gene-flow across the

once panmictic population may result in higher degrees of fixation in regions of the genomes

of bats from different geographic areas.

153   MATERIALS AND METHODS

Ethics Statement

This study was carried out on bats from non-endangered species in strict accordance with

recommendations in the Guide for the Care and Use of Laboratory Animals of the National

Institutes of Health. All methods were approved by the Institutional Animal Care and Use

Committee at Bucknell University (protocols DMR-016 and DMR-17). The bats were

collected under Pennsylvania Game Commission Special Use Permit #33085, State of

Michigan Scientific Collector's Permit #1475, and New York State Department of

Environmental Conservation Permit #427.

Sample collection and DNA extraction

We conducted whole-genome sequencing on a population of *M. lucifugus* prior to, and up to 10 years after the onset of WNS using a total of 219 samples (Table 1.). For historic samples, wing tissue for sequencing was obtained from frozen specimens of known origin (Supp. Table 1.) that were collected and stored by the New York State Department of Environmental Conservation (Delmar, NY) and by D.M. Reeder at Bucknell University in Pennsylvania (PA). White-nose syndrome in bats was first observed in upstate New York (NY), at Howe's Cave during the winter of 2005-2006 (Blehert *et al.* 2009). Our historic samples were obtained from individuals from of the eastern population of *M. lucifugus* (Miller-Butterworth *et al.* 2014; Vonhof *et al.* 2015) from New York (NY) (Miller-Butterworth *et al.* 2014; Vonhof *et al.* 2015), including individuals from the first batch of dead bats found at Hailes Cave, NY, during the winter of 2006-2007 and from animals caught in central PA in 2005 and 2006 (Fig. 1., Table 1., Supp. Table 1.). Although the bat originally described as *M. lucifugus* in North America has since been divided into five non-sister species (Morales and Carstens 2018), all of the individuals sampled in our study belong to *M. lucifugus senso stricto*, with a range extending from the east coast of North America to Alaska, and furthermore, to the same previously assigned population (Vonhof *et al.* 2015; Wilder *et al.* 2015). Previous nuclear genetic studies show that differentiation is low, and there is no evidence for any major barriers to nuclear gene flow across the eastern range of *M. lucifugus* (Vonhof *et al.* 2015; Wilder *et al.* 2015). The bats from which the historic samples in NY were collected were affected by WNS at the time of collection. However, because they were amongst the first bats to be documented with the disease in North America, we believe this set of samples is representative of the population genetic structure of *M. lucifugus* prior to

WNS. Bats in PA became affected by the disease in the winter of 2008–2009 and thus had

not been affected by WNS at the time of sampling in 2006-2007. Samples from PA and NY

in 2006-2007 are therefore called 'pre-WNS' hereafter.

The remnants of the eastern North American population in NY and PA were sampled again in

2015-2016, and genetic diversity was compared before and after WNS (Fig.1., Table 1.,

Supp. Table 1.). Post-WNS bats were captured using mist nets and harp traps during 2015 -

2016 from a number of maternal colonies and hibernacula. In addition to samples from PA

and NY, we collected gDNA from two individual bats from Upper-Peninsula Michigan

(UPMI) in 2014 for use in polymorphism detection. We did this to take advantage of the

greater number of polymorphic sites one would expect to arise from a more diverse collection

of individuals (see sequencing methods, below). Tissue samples were collected using either

3.0 mm or 5.0 mm biopsy punches (Integra Milltex, Plainsboro, NJ) and stored in 95%

ethanol until extraction. We extracted DNA using QIAmp DNA Mini Kits (Qiagen, Hilden

Germany) and stored DNA at -80°C until sequencing.

Sequencing methods

Sequencing libraries were made for two separate sets of data: **(1)** Six individuals from PA (N

= 2), NY (N = 2) and UPMI (N= 2) were sequenced separately in order to initially identify a

set of polymorphic sites within the *M. lucifugus* genome within the focal population (See

Supp. Table 1. for information on individuals). Individual sequencing increases the

confidence of SNP discovery, relative to pooled sequencing, since individuals exhibit discrete

variation in the number of alleles they carry at each potential SNP site (Schlötterer *et al.*

2014); **(2)** Equimolar pools of DNA from multiple individuals from within our north-eastern

US population (NY and PA) at each of the two timepoints (pre- and post-WNS) were

combined to give four pooled sequencing libraries (See Table 1. and Supp. Table 1. for

information on individuals). Without *a priori* knowledge of population structuring within our

sampling area (although see Miller-Butterworth et al., 2014 for PA), we divided our samples

in two according to geographic distance. This produced a division between the political

boundaries of PA and NY. TruSeq Nano libraries with a 350 bp insert size were prepared

from all samples and run on a HiSeq4000 platform to generate 2x 150 bp reads providing ~15

Gbp of sequence data (~7.5x coverage) for each of the six individuals **(1)** and ~90 Gbp of

sequence data (~45x coverage) for each of the four pools **(2)**. Adapter sequences were

removed with Cutadapt v1.2.1 (Martin 2011) and trimmed with Sickle 1.200 (Joshi and Fass

2011) with a minimum quality score of 20.

Bioinformatics methods

First, reads from the six individually sequenced individuals were mapped onto the *M.*

lucifugus reference genome (https://www.broadinstitute.org/brown-bat/little-brown-bat-

genome-project) using Bowtie2 v2.2.5 (Langmead and Salzberg 2012), with the '—sensitive-

local' preset. Reads were sorted using SAMtools v1.1 (Li *et al.* 2009) and combined by read

group (with duplicates removed using a 'lenient' validation stringency setting) using Picard-

tools v2.5.0 (http://broadinstitute.github.io/picard). SNPs were called across all six

individuals with the Genome Analysis Toolkit (GATK) v3.4 HaplotypeCaller (McKenna *et*

al. 2010) using default parameters. Within this SNP set, high-quality bivalent SNPs were

chosen that met selection criteria across the six individuals (Quality Score > 100; AF<1;

30>=DP>=100) and were input to the SAMtools mpileup function as a list of sites at which

pileup was to be generated. After SNP calling, these six individuals were discarded from

further analyses.

Reads from the four pooled samples were mapped to the *M. lucifugus* reference genome as

above and SAMtools mpileup was used to count alleles at high-quality SNP sites identified

above from the six sequenced individuals (i.e. reads matching either the reference or alternate

allele for bivalent loci). Sites with coverage in the bottom 10% or top 10% quantiles summed

across the pooled samples were excluded (because these could reflect inaccuracies in the

reference genome), leaving approximately 13.5 million sites.

Average heterozygosity within the pooled samples was calculated using npstats (Ferretti *et al.*

2013), which controls for number of individuals within each sequencing pool and sequencing

depth, thus allowing robust comparisons between pooled samples. One million SNPs were

randomly picked and permuted, with 1000 sets of 1000 SNPs used to calculate the observed

heterozygosity within each pooled set of individuals and, from these, we produced the

median and 95% confidence intervals on heterozygosity.

To detect regions of the genome that may be under selection, we plotted F_{ST} across the *M.*

lucifugus genome. Due to our sampling and sequencing regime, we were also able to

separately look at individuals of the same overall population from two geographical locations

within their range. In our case, an arbitrary divide was made between bats sampled in NY and

PA (Figure 1), dividing our sampling population roughly at the midline. Although population

genetic analysis based on a considerably lower number of genetic markers assigns *M.*

lucifugus on the eastern coast of North America to the same population (Vonhof *et al.* 2015),

the species shows high degrees of philopatry (Davis and Hitchcock 1965; Norquay *et al.*

2013), warranting the examination of these arbitrarily assigned pools of individuals outside

the mean dispersal range of the species using more accurate, whole-genome sequencing

based population genetic analyses. Using allele counts at each locus to estimate allele

frequencies, F_{ST} was calculated without weighting for sampling effort

$(F_{ST} = (\bar{p}(1-\bar{p}) - \overline{p(1-p)})/\bar{p}(1-\bar{p})$, where p is the allele frequency and either

averaged between populations $(\bar{p})$ or the value $p(1-p)$ averaged across populations). F_{ST}

was separately calculated between the Pre-WNS and Post-WNS samples and between

subsections thereof, specifically PA pre- vs post-WNS, NY pre- vs post-WNS, PA pre- vs

NY pre-WNS, PA post- vs NY post-WNS, and all PA vs all NY). Selection is expected to

elevate F_{ST} at specific genomic regions relative to background levels due to sampling noise or

genetic drift, which are expected to act equally across the genome (Holsinger and Weir

2009). To identify such regions across the genome, a moving window was applied to

calculate median pairwise F_{ST} values for groups of 500 consecutive SNPs at a time to even

out sampling effects. Moving windows shifted 100 SNPs at a time, thus providing the best

balance between fine detail and practicality. Allele frequency estimates were weighted by the

number of individuals in each pool. This was done using a custom R script (see Data

availability section), run using R v3.4.1 (R Development Core Team 2011). Results were

confirmed using Popoolation2 with a 10 kb sliding window and with defined pool sizes. This

gave the same qualitative results, but higher levels of background noise (Kofler *et al.* 2011).

Additionally, Popoolation2 was used to perform Fishers Exact Tests on individual SNPs in

order to test whether any very highly significant SNPs were missed by the moving window

approach.

From the comparison of the NY and PA pooled samples (i.e. the comparison that produced

the highest F_{ST} values), we visually identified, across the whole genome, windows with

median F_{ST} values greater than 5 standard deviations from the mean, as well as those with F_{ST}

values greater than 0.05, indicating at least 'moderate' genetic differentiation (Wright, 1978).

The numbers of such complete windows within gene models were counted for each gene to

identify genes potentially under selective pressure.

Data availability

Raw data are available at ENA accession ERP120680, and bioinformatics code at

https://github.com/scottishwormboy/myoluc.

282   RESULTS

Pre-WNS vs Post-WNS comparisons across all sampled bats

Comparing F_{ST} between Pre- and Post-WNS samples revealed an overall low level of

population differentiation due to genetic structure (F_{ST} median of 0.0059; Table 2). While a

small number of windows have F_{ST} values slightly beyond our arbitrary cut-off of 5 standard

deviations above the mean, none have F_{ST} values exceeding 0.05, suggesting that there is not

even moderate selection in any region of the genome in Post-WNS samples compared to Pre-

WNS samples (Fig. 2, Table 2)(Wright 1978). Similarly, using Fisher exact tests from

Popoolation2 at a threshold of $p < 10^{-6}$, no individual SNPs were found to be consistently

associated with differences Pre- versus Post-WNS among the four sets of samples.

Comparisons between bat populations in different geographic locations

Data split by geographic location (PA versus NY) revealed a number of minor differences ($F_{ST} > 5$ stdevs above mean, but $F_{ST} < 0.05$) by geographic location prior to arrival of WNS, as well as a peak in scaffold GL429835 with an F_{ST} value exceeding 0.05 (Fig. 3A). The high peak consists of 4 windows, the midpoints of which span an 81,887 bp region within a coding sequence consisting of 47 exons, which encodes the NIPBL cohesion loading factor (NCBI Gene ID: 102430106). The highest peak in a comparison of PA populations Pre- and Post-WNS (Fig. 3B) lies in a similar position within this coding region, as does a peak in a similar comparison of NY samples (Fig. 3C), though neither exceed the F_{ST} value cut-off of 0.05.

When comparing the Post-WNS individuals sampled from PA with the Post-WNS individuals from NY, the peak on scaffold GL429835 is no longer visible, but a far more pronounced spike in F_{ST} values can be clearly seen in scaffold GL429776 (Fig. 4A), indicating relatively large allele frequency differences between the sample sets in this region of the genome. This is also seen, to a lesser extent, when comparing all samples from NY with all samples from PA (Fig. 4B). However, this significant peak in this comparison of all samples collected does not affect average fixation indices across the genome relative to the other comparison of all samples, Pre-WNS vs Post-WNS (Table 2). This peak could be attributed to mapping error, but our coverage depths show an average depth of 66.33x across the whole genome, whereas the focal region was sequenced at an average depth of 66.49x. This means that there has been no collapsing of duplicated regions, nor poor coverage of the region. Also, viewing the data on the NCBI genome viewer tool, with RepeatMasker

histogram enabled, suggested no immediately obvious increase in repetitive sequence in the

region of interest.

We examined scaffold GL429776 more closely to compare the F_{ST} values between sample

sets from NY Post-WNS and PA Post-WNS (Fig. 4A). Genes within scaffold GL429776

containing windows with fixation indices > 0.05 between the NY Post-WNS and PA Post-

WNS samples, are presented in Table 4 and Fig. 4A. To clarify why we saw no fixation at

this site when comparing Pre-WNS samples from both geographic locations (Fig. 3A), nor

when we separately compared Pre-WNS and Post-WNS populations from each geographic

location (Figs. 3B and 3C), we calculated the proportion of reference alleles called at each

site for bats from each geographic area included in the F_{ST} window analyses (Fig. 5). This

revealed a small decrease over time in the proportion of the reference alleles seen at this

locus in the PA population (Fig. 5A), commensurate with an increase in proportion of

reference alleles in the NY sample set (Fig. 5B). As such, existing differences between the

NY and PA sample sets were insufficient to cause a relative rise in fixation indices prior to

WNS spreading through the population (Fig 5C). Conversely, Post-WNS, allele proportions

at this site in both sample sets had not changed sufficiently to suggest selection *within* the

sample set but demonstrated clear differences *between* the NY and PA sample sets (Fig. 5D).

In the Post-WNS comparison, peaks that were above our arbitrary threshold of 5 standard

deviations above the mean across the whole genome, but below $F_{ST} = 0.05$, can also be

observed in small scaffolds on the right-hand side of the F_{ST} plots (Fig. 4A). The 63 windows

comprising these peaks lie in 47 scaffolds. Of these, high F_{ST} windows lie in only 6 protein-

coding regions across 6 scaffolds (Table 3). Two of these encode partial gene models. Of the

4 remaining models containing high F_{ST} regions, all but 1 is described on NCBI as a "Low

Quality Protein". The exception is a gene encoding the ankyrin 2 protein, within which 7 high

F_{ST} windows lie.

Measures of heterozygosity within geographically and temporally-distinct sample sets

340   We examined heterozygosity across the genome for each geographic location at each

341   timepoint to measure whether the epizootic reduced genetic diversity. No change in genetic

342   diversity was detected in either geographic location: PA pre-WNS: Median: 0.263 (0.252 –

343   0.274 95% CI); PA post-WNS: 0.247 (0.239 – 0.255); NY pre-WNS: 0.255 (0.247 – 0.263);

344   and NY post-WNS: 0.256 (0.247 -0.265).

345   DISCUSSION

Examining our data as a whole, we found low F_{ST} values between the Pre- and Post-WNS

individuals of our eastern North American *M. lucifugus* population. Also, we found no

significant changes in within-population heterozygosity between Pre- and Post-WNS samples

in either PA or NY, despite the massive population declines caused by WNS over the past

decade, and found overall high heterozygosity across the genome. Together, these data

suggest that changes in population size due to WNS have not been sufficient to affect genetic

structure or diversity between time points.

Geographic separation of our bats down the middle of our total sampling region allowed us to

identify a region of the genome that showed differentiation between Post-WNS PA and Post-

WNS NY samples. A closer examination revealed that differentiation at this region was

present already prior to WNS, but fixation in the PA sample set had increased between our

two temporal sampling points. Here, we see low F_{ST} values between the two groups of bats

prior to the onset of WNS, indicating an absence of selection across the genome, with the

exception of a small peak in scaffold GL429835 (Fig. 3A), which lies within a gene model

encoding the NIPBL cohesion loading factor. Somewhat elevated F_{ST} values are also visible

within this region when comparing the PA population before and after the spread of WNS, as

well as the NY population pre- and post-WNS (Figs. 3B and 3C). However, this peak is not

evident when comparing PA and NY post-WNS (Fig. 4a). Any role of spatio- or temporal

selection on this region is therefore weak and unlikely to be related to WNS. The region

encodes the protein NIPBL, which plays an important role in the function of the cohesin

complex, which is vital for chromosome segregation, DNA damage repair, and gene

expression regulation (Peters *et al.* 2008). What role variation in this region could play in

natural populations is unknown.

Our overall results confirm that the pre-WNS bats from NY were genetically similar to pre-

WNS bats from PA. We should note, however, that it is possible that our NY samples

collected from dead bats at Hailes Cave in February 2007 (Site 5, Fig. 1, Supp. Table. 1), a

year after WNS was first detected at a single site in NY (Blehert et al., 2009), may not fully

represent the genetic composition of the Pre-WNS NY bat population and that our analysis

could have benefitted from the sequencing of samples from a variety of sites. Regarding Post-

WNS sampling; because individual maternal colonies and hibernacula may show a higher

degree of relatedness between individuals (Kurta and Murray 2002; Burns *et al.* 2014;

Olivera-Hyde *et al.* 2019), we included bats from a number of these sites in the study.

However, summer colonies and overwintering sites may be separated by considerable

distances and therefore there will always be some degree of gene flow at the geographical

scales considered, i.e., among sampling sites within a state (Norquay *et al.* 2013).

The F_{ST} comparison of Post-WNS individuals sampled from PA and NY depicts considerable

genetic differentiation within scaffold GL429776 (Fig. 4A); differentiation which did not

exist in the Pre-WNS comparison of individuals sampled from these geographical areas. This

is a difference that would have gone unnoticed had we not made the comparison between bats

sampled in NY and PA, or used fewer genetic markers instead of a whole-genome approach

(Miller-Butterworth *et al.* 2014; Vonhof *et al.* 2015). The appearance post-WNS of this

region of greater genetic differentiation within scaffold GL429776 when considering all

geographical locations together is not, however, reflected when examining temporal changes

in either geographic location on their own (Fig. 3B and Fig. 3C). A comparison of allele

frequencies of individual SNPs within this region shows that individuals sampled from PA

showed an increase in deviation from the reference genome over time, while a greater

proportion of reference alleles were seen in NY samples Post-WNS compared with Pre-

WNS. The differences in proportions, growing as they were over time, were insufficient to

exhibit a high fixation index Pre-WNS, but were notably different after the disease had

spread through the population. One could argue that this is due to sampling at maternal

colonies and swarming sites, which has known effects on allele frequency differences

(Johnson *et al.* 2015). However, we sampled at a number of maternal colonies, which would

dilute this effect, and furthermore, recent studies suggest philopatry within maternal colonies

may not be as high as previously expected (Olivera-Hyde et al., 2019, although on *M.*

septentrionalis). Therefore, we assume our sampling of multiple maternal colonies and

hibernation sites has removed any effects of relatedness within our sample sets.

The region within the scaffold GL429776 containing the peak has six annotated genes with

very different functions (The UniProt Consortium 2019). Two, POC1B and CEP290, are

involved in cilia development and maintenance. The ATP2B1 gene encodes a protein which

belongs to the family of P-type primary ion transport ATPases, whereas TMTC3 is involved

in the control of endoplasmic reticulum stress response. MGAT4C is a glycosyltransferase

that participates in the transfer of N-acetylglucosamine (GlcNAc) to the core mannose

residues of N-linked glycans. Finally, KITLG produces a protein involved in mast cell

development, migration and function, and melanogenesis. KITLG has been documented as a

target of evolution in recent studies and in humans it has been linked with thermoregulation

(Yang *et al.* 2018), which is a critical component of bat life history (Studier and O'Farrell

1972). In fact, upregulation of KITLG at low temperature helps promote the production of

brown fat for heat generation (Huang *et al.* 2014). Bats utilise brown adipose tissue in

arousals to normothermia from different degrees of torpor (Thomas *et al.* 1990). There is also

evidence of parallel evolution of pigmentation in sticklebacks and humans linked to KITLG

(Miller *et al.* 2007), as well as its role in determining resistance/susceptibility to swine

respiratory diseases in Erhulian pigs (Huang *et al.* 2017). KITLG, by influencing mast cell

development, could also play a role in either protective or pathological immune responses to

P. destructans, which may involve IgE-mediated recognition of secreted fungal proteins

(Reeder *et al.* 2017; Field *et al.* 2018). As such, temporal changes we found in allele

frequencies at this locus could, for instance, be linked to differences in hibernation site

temperatures, and/or thermoregulatory needs in bats sampled at different locations (Thomas

et al. 1990; Humphries *et al.* 2002). Spatial patterns, as seen here between the arbitrarily

pooled samples from PA and NY, could be due to local adaptation, in which strong selective

sweeps may be linked to gene variants favored in local interactions (Hansen *et al.* 2012;

Schoville *et al.* 2012; Kyle *et al.* 2014; Rico *et al.* 2015). We lack data concerning migration

patterns of the bats studied here, as well as the environmental conditions within their

microfugia. Future work would benefit from these data in a bid to understand whether bats

repeatedly frequent habitats with similar conditions and whether there is a significant

difference in conditions across these eastern states that explains the rapid rise in allelic

differentiation, and thus F_{ST} values, at this locus.

In contrast to our study, in which we did not detect evidence for an evolutionary rescue effect

at a large geographic scale, Auteri & Knowles (2020)and Donaldson et al. (2017) found

putative selectively driven genetic changes in local populations of *M. lucifucus*, which could

have the potential to result in an evolutionary rescue from WNS. Differences in results may

have arisen among the studies because of the more limited sampling or lack of repeated

sampling before and after WNS in the other two studies, or because of the different

geographic areas sampled. Further, the Donaldson et al. (2017) study presented subtle

immunogenetic variation across a wide geographic area, with the post-WNS population

sampled in the first or second winter of WNS exposure. Thus, differences among populations

in that study could represent local adaptation, independent of the effects of WNS. An

alternate possibility that could explain the findings of all three studies is that gene flow in the

formerly panmictic eastern population of *M. lucifugus* could have been reduced following

massive population declines, and adaptive responses to WNS could be beginning to emerge

independently in localized areas, even as broad-scale adaptation is not evident across the

range of the species. Further clarification of the role of the genes identified in each of the

three studies could be revealed by studying differences in transcriptomes, infection, and

survival in response to WNS during infection trials in hibernation experiments (Field *et al.*

2018; Lilley *et al.* 2019).

Despite not seeing an overarching genetic signal associated with resistance in *M. lucifugus*,

many remnant populations appear to be surviving year after year and in some cases even

increasing in size (Langwig et al., 2017; Lilley et al., 2016). The pathogen loads appears to be

significantly lower for remnant populations compared to that of bats during the epidemic

phase, during which massive declines were, which would suggest the development of

resistance in these populations (Bernard et al., 2017; Langwig et al., 2017; but see Lilley et

al., 2019). However, improved responses to WNS may occur through other means than the

evolution of resistance. For instance, during the hibernation period, when the bat hosts are

most vulnerable to the pathogen, roost microclimate variables, including humidity and

temperature, affect the growth of the fungal pathogen and the survival of the host (Verant *et*

al. 2012; Johnson *et al.* 2014; Grieneisen *et al.* 2015; Marroquin *et al.* 2017), and behavioral

shifts in roost site selection by bat hosts since the onset of WNS (Johnson *et al.* 2016) could

favor energy conservation while reducing fungal growth and infection. Modelling also

suggests that environmental factors, including the cave microbiome, have an impact on the

proliferation and infectivity of *P. destructans* (Hayman, Pulliam, Marshall, Cryan, & Webb,

2016; Lilley et al., 2018). This is supported by the discovery of microbes in hibernacula

environments and on bats that are able to retard the growth of the fungus (Zhang *et al.* 2014,

2015; Micalizzi *et al.* 2017). In light of our results and considering bat life history as a whole,

adaptation and evolutionary rescue may not be a fast track for recovery of bat populations

affected by WNS (Maslo and Fefferman 2015), although ultimately these will contribute to

long term survival of populations (Lilley *et al.* 2019). At the present, therefore, it is more

likely that behavioral shifts in selection of hibernation sites that vary in environmental

conditions and strong selection for microbial taxa that inhibit *P. destructans* could explain

why some colonies have persisted or may have even begun to recover from the zoonosis

(Cheng *et al.* 2016; Johnson *et al.* 2016; Lemieux-Labonté *et al.* 2017; Lilley *et al.* 2018).

More broadly, it is still unclear how frequently genetic adaptation occurs in natural

populations and under what circumstances it is promoted (Schoville *et al.* 2012). Many

studies have recorded phenotypic changes in natural populations and attributed them to host-

pathogen interactions (Kilpatrick 2006; Råberg *et al.* 2009; Frank *et al.* 2014; Langwig *et al.*

2017). However, it is difficult to determine conclusively whether changes in phenotype are

the result of selection in the genome or a result of phenotypic plasticity (Paterson *et al.* 2010;

Routtu and Ebert 2015). Also, the assumed benefit of change (i.e. adaptation) is often not

tested experimentally, which would allow for inference of causality with the pathogen

through exclusion of other potential drivers.

To conclude, our results suggest WNS has not yet decidedly subjected populations of *M.*

lucifugus to selective pressures leading to genetic adaptation in remnant populations. No

population -wide signs of selection in comparisons of genomes of Pre-WNS and ten years

Post-WNS populations were observed in this study (although see Donaldson *et al.* 2017;

Auteri and Knowles 2020). This is something that is not unexpected with a species in which

the effective population size most likely numbered in tens of millions of individuals prior to

the onset of WNS. The disease has not caused a true population bottleneck, as exemplified by

our measures of genetic diversity not varying between pre- and post-WNS samples. Our

results indicate that the persistence and recent growth of some remnant populations of *M.*

lucifugus are more likely attributable to other factors, such as microbiome adaptation and

hibernation site selection among others, rather than genetic adaption. However, we found

increased variability in a specific area of the genome in a comparison of Post-WNS samples

from our two different geographic locations, relative to Pre-WNS samples. We suggest this is

due to weakened connectivity between bats at different locations allowing for local

adaptations to appear in the absence of gene flow. Although genes in the high F_{ST} region of

the genome were identified, and genes such as KITLG provide interesting avenues of

research, even with regards to WNS, further investigation into the processes in bat life history

to which these genes are related are required to expound upon the existing gene annotations.

502   ACKNOWLEDGEMENTS

We thank Svenska Kulturfonden, H2020 Marie Skłodowska☐Curie Actions (TML, 706196)

and the Natural Environment Research Council, UK for funding the study; staff at the Centre

for Genomic Research, University of Liverpool, UK for DNA sequencing expertise; Jenni

Prokkola and Riley F. Bernard for proofreading the manuscript; and Alice Lilley, Cali

Wilson, Beth Rogers, Mike Scafini, Scott Wasilko, Cassandra Ostroski, Heather Rogers,

Emily Blackman, Kristin Gaines, Rachel Mosely, Angela Remeika, Imran Ejotre, Laura

Kurpiers, Joseph S. Johnson, Marianne Gagnon, Spencer Schell, Sarah Bouboulis, Karen El

Chaar and Allentown Parks for assisting in sample collection.

511   LITERATURE CITED

Auteri, G. G., and L. L. Knowles, 2020 Decimated little brown bats show potential for

adaptive change. Sci. Rep. 10: 1–10.

Bernard, R. F., and G. F. McCracken, 2017 Winter behavior of bats and the progression of

white☐nose syndrome in the southeastern United States. Ecol. Evol. 7: 1487–1496.

Bernard, R. F., E. V. Willcox, K. L. Parise, J. T. Foster, and G. F. McCracken, 2017 White-

nose syndrome fungus, *Pseudogymnoascus destructans*, on bats captured emerging

from caves during winter in the southeastern United States. BMC Zool. 2: 12.

Blehert, D. S., A. C. Hicks, M. Behr, C. U. Meteyer, B. M. Berlowski-Zier *et al.*, 2009 Bat

white-nose syndrome: an emerging fungal pathogen? Science 323: 227–227.

Boots, M., A. Best, M. R. Miller, and A. White, 2009 The role of ecological feedbacks in the

evolution of host defence: what does theory tell us? Philos. Trans. R. Soc. Lond. B

Biol. Sci. 364: 27–36.

Brownlee-Bouboulis, S. A., and D. M. Reeder, 2013 White-nose syndrome-affected little

brown myotis (*Myotis lucifugus*) increase grooming and other active behaviors during

arousals from hibernation. J. Wildl. Dis. 49: 850–859.

Burns, L. E., T. R. Frasier, and H. G. Broders, 2014 Genetic connectivity among swarming

sites in the wide ranging and recently declining little brown bat (*Myotis lucifugus*).

Ecol. Evol. 4: 4130–4149.

Campana, M., N. Kurata, J. T. Foster, L. Helgen, D. Reeder *et al.*, 2017 White-nose

syndrome fungus in a 1918 bat specimen from France. Emerg. Infect. Dis. J. 23:

1611–1612.

Campbell, M. C., and S. A. Tishkoff, 2008 African genetic diversity: Implications for human

demographic history, modern human origins, and complex disease mapping. Annu.

Rev. Genomics Hum. Genet. 9: 403–433.

Cheng, T. L., A. Gerson, M. S. Moore, J. D. Reichard, J. DeSimone *et al.*, 2019 Higher fat

stores contribute to persistence of little brown bat populations with white-nose

syndrome. J. Anim. Ecol. 88: 561–600.

Cheng, T. L., H. Mayberry, L. P. McGuire, J. R. Hoyt, K. E. Langwig *et al.*, 2016 Efficacy of

a probiotic bacterium to treat bats affected by the disease white-nose syndrome. J.

Appl. Ecol. 54: 701–708.

Cryan, P. M., C. U. Meteyer, D. S. Blehert, J. M. Lorch, D. M. Reeder *et al.*, 2013 Electrolyte

depletion in white-nose syndrome bats. J. Wildl. Dis. 49: 398–402.

Cryan, P. M., C. U. Meteyer, J. G. Boyles, and D. S. Blehert, 2010 Wing pathology of white-

nose syndrome in bats suggests life-threatening disruption of physiology. Bmc Biol.

8: 135.

Daszak, P., L. Berger, A. A. Cunningham, A. D. Hyatt, D. E. Green *et al.*, 1999 Emerging

infectious diseases and amphibian population declines. Emerg. Infect. Dis. 5: 735–

748.

Daszak, P., A. A. Cunningham, and A. D. Hyatt, 2000 Emerging infectious diseases of

wildlife-threats to biodiversity and human health. Science 287: 443–449.

Davis, W. H., and H. B. Hitchcock, 1965 Biology and migration of the bat, *Myotis lucifugus*,

in New England. J. Mammal. 46: 296–313.

Davy, C. M., M. E. Donaldson, Y. Rico, C. L. Lausen, K. Dogantzis *et al.*, 2017 Prelude to a

panzootic: Gene flow and immunogenetic variation in northern little brown myotis

vulnerable to bat white-nose syndrome. FACETS 2: 690–714.

De Castro, F., and B. Bolker, 2005 Mechanisms of disease-induced extinction. Ecol. Lett. 8:

117–126.

Deschamps, M., G. Laval, M. Fagny, Y. Itan, L. Abel *et al.*, 2016 Genomic signatures of

selective pressures and introgression from archaic hominins at human innate

immunity genes. Am. J. Hum. Genet. 98: 5–21.

Di Gaspero, G., D. Copetti, C. Coleman, S. D. Castellarin, R. Eibach *et al.*, 2012 Selective

sweep at the Rpv3 locus during grapevine breeding for downy mildew resistance.

TAG Theor. Appl. Genet. Theor. Angew. Genet. 124: 277–286.

Dobony, C. A., and J. B. Johnson, 2018 Observed resiliency of little brown myotis to long-

term white-nose syndrome exposure. J. Fish Wildl. Manag. 9: 168–179.

Donaldson, M. E., C. M. Davy, C. K. R. Willis, S. McBurney, A. Park *et al.*, 2017 Profiling

the immunome of little brown myotis provides a yardstick for measuring the genetic

response to white-nose syndrome. Evol. Appl. 10: 1076–1090.

Ferretti, L., S. E. Ramos-Onsins, and M. Pérez-Enciso, 2013 Population genomics from pool

sequencing. Mol. Ecol. 22: 5561–5576.

Field, K. A., J. Johnson, T. Lilley, S. Reeder, E. Rogers *et al.*, 2015 The white-nose

syndrome transcriptome: activation of anti-fungal host responses in wing tissue of

hibernating bats. Plos Pathog. 11: e1005168.

Field, K. A., B. J. Sewall, J. M. Prokkola, G. G. Turner, M. F. Gagnon *et al.*, 2018 Effect of

torpor on host transcriptomic responses to a fungal pathogen in hibernating bats. Mol.

Ecol. 27: 3727–3743.

Frank, C. L., A. D. Davis, and C. Herzog, 2019 The evolution of a bat population with white-

nose syndrome (WNS) reveals a shift from an epizootic to an enzootic phase. Front.

Zool. 16: 40.

Frank, C. L., A. Michalski, A. A. McDonough, M. Rahimian, R. J. Rudd *et al.*, 2014 The

resistance of a North American bat species (*Eptesicus fuscus*) to white-nose syndrome

(WNS) (M. L. Baker, Ed.). PLoS ONE 9: e113958.

Frick, W. F., T. L. Cheng, K. E. Langwig, J. R. Hoyt, A. F. Janicki *et al.*, 2017 Pathogen

dynamics during invasion and establishment of white-nose syndrome explain

mechanisms of host persistence. Ecology 98: 624–631.

Frick, W. F., J. F. Pollock, A. C. Hicks, K. E. Langwig, D. S. Reynolds *et al.*, 2010 An

emerging disease causes regional population collapse of a common North American

bat species. Science 329: 679–682.

Frick, W. F., S. J. Puechmaille, J. R. Hoyt, B. A. Nickel, K. E. Langwig *et al.*, 2015 Disease

alters macroecological patterns of North American bats. Glob. Ecol. Biogeogr. 24:

741–749.

Gallana, M., M.-P. Ryser-Degiorgis, T. Wahli, and H. Segner, 2013 Climate change and

infectious diseases of wildlife: Altered interactions between pathogens, vectors and

hosts. Curr. Zool. 59: 427–437.

Grieneisen, L. E., S. A. Brownlee-Bouboulis, J. S. Johnson, and D. M. Reeder, 2015 Sex and

hibernaculum temperature predict survivorship in white-nose syndrome affected little

brown myotis (Myotis lucifugus). R. Soc. Open Sci. 2: 140470.

Groot, N. G. de, N. Otting, G. G. M. Doxiadis, S. S. Balla-Jhagjhoorsingh, J. L. Heeney *et*

al., 2002 Evidence for an ancient selective sweep in the MHC class I gene repertoire

of chimpanzees. Proc. Natl. Acad. Sci. 99: 11748–11753.

Hansen, M. M., I. Olivieri, D. M. Waller, and E. E. Nielsen, 2012 Monitoring adaptive

genetic responses to environmental change. Mol. Ecol. 21: 1311–1329.

Harazim, M., I. Horáček, L. Jakešová, K. Luermann, J. C. Moravec *et al.*, 2018 Natural

selection in bats with historical exposure to white-nose syndrome. BMC Zool. 3: 8.

Hayman, D. T. S., J. R. C. Pulliam, J. C. Marshall, P. M. Cryan, and C. T. Webb, 2016

Environment, host, and fungal traits predict continental-scale white-nose syndrome in

bats. Sci. Adv. 2: e1500831.

Holsinger, K. E., and B. S. Weir, 2009 Genetics in geographically structured populations:

defining, estimating and interpreting FST. Nat. Rev. Genet. 10: 639–650.

Huang, X., T. Huang, W. Deng, G. Yan, H. Qiu *et al.*, 2017 Genome-wide association studies

identify susceptibility loci affecting respiratory disease in Chinese Erhualian pigs

under natural conditions. Anim. Genet. 48: 30–37.

Huang, Z., H.-B. Ruan, L. Xian, W. Chen, S. Jiang *et al.*, 2014 The stem cell factor/Kit

signalling pathway regulates mitochondrial function and energy expenditure. Nat.

Commun. 5: 4282.

Humphries, M. M., D. W. Thomas, and J. R. Speakman, 2002 Climate-mediated energetic

constraints on the distribution of hibernating mammals. Nature 418: 313–316.

Ingersoll, T. E., B. J. Sewall, and S. K. Amelon, 2016 Effects of white-nose syndrome on

regional population patterns of three hibernating bat species. Conserv. Biol. 30: 1048–

1059.

Ingersoll, T. E., B. J. Sewall, and S. K. Amelon, 2013 Improved analysis of long-term

monitoring data demonstrates marked regional declines of bat populations in the

eastern United States. PLOS ONE 8: e65907.

Jeffery, K. J., and C. R. Bangham, 2000 Do infectious diseases drive MHC diversity?

Microbes Infect. 2: 1335–1341.

Johnson, L. N. L., B. A. McLeod, L. E. Burns, K. Arseneault, T. R. Frasier *et al.*, 2015

Population genetic structure within and among seasonal site types in the little brown

bat (*Myotis lucifugus*) and the northern long-eared bat (*M. septentrionalis*). PLOS

ONE 10: e0126309.

Johnson, J. S., D. M. Reeder, J. W. McMichael, M. B. Meierhofer, D. W. F. Stern *et al.*, 2014

Host, pathogen, and environmental characteristics predict white-nose syndrome

mortality in captive little brown myotis (*Myotis lucifugus*). Plos One 9: e112502.

Johnson, J. S., M. Scafini, B. J. Sewall, and G. G. Turner, 2016 Hibernating bat species in

Pennsylvania use colder winter habitats following the arrival of white-nose syndrome.

Editors: Calvin M. Butchkoski, DeeAnn M. Reeder, Gregory G. Turner, Howard P.

Whidden, pp. 181–199 in *Conservation and Ecology of Pennsylvania's Bats,* The

Pennsylvania Academy of Sciences.

Joshi, N., and J. Fass, 2011 Sickle: A sliding-window, adaptive, quality-based trimming tool

for   FastQ   files   (Version   1.33)   [Software].   Available   at

https://github.com/najoshi/sickle.

Karlsson, E. K., D. P. Kwiatkowski, and P. C. Sabeti, 2014 Natural selection and infectious

disease in human populations. Nat. Rev. Genet. 15: 379–393.

Kilpatrick, A. M., 2006 Facilitating the evolution of resistance to avian malaria in Hawaiian

birds. Biol. Conserv. 128: 475–485.

Kofler, R., R. V. Pandey, and C. Schlötterer, 2011 PoPoolation2: identifying differentiation

between populations using sequencing of pooled DNA samples (Pool-Seq).

Bioinformatics 27: 3435–3436.

Kurta, A., and S. W. Murray, 2002 Philopatry and migration of banded Indiana bats (*Myotis*

sodalis) and effects of radio transmitters. J. Mammal. 83: 585–589.

Kyle, C. J., Y. Rico, S. Castillo, V. Srithayakumar, C. I. Cullingham *et al.*, 2014 Spatial

patterns of neutral and functional genetic variations reveal patterns of local adaptation

in raccoon (*Procyon lotor*) populations exposed to raccoon rabies. Mol. Ecol. 23:

2287–2298.

Langmead, B., and S. L. Salzberg, 2012 Fast gapped-read alignment with Bowtie 2. Nat.

Methods 9: 357–359.

Langwig, K. E., W. F. Frick, R. Reynolds, K. L. Parise, K. P. Drees *et al.*, 2015 Host and

pathogen ecology drive the seasonal dynamics of a fungal disease, white-nose

syndrome. Proc. R. Soc. B-Biol. Sci. 282: 20142335.

Langwig, K. E., J. R. Hoyt, K. L. Parise, W. F. Frick, J. T. Foster *et al.*, 2017 Resistance in

persisting bat populations after white-nose syndrome invasion. Phil Trans R Soc B

372: 20160044.

Lemieux-Labonté, V., A. Simard, C. K. R. Willis, and F.-J. Lapointe, 2017 Enrichment of

beneficial bacteria in the skin microbiota of bats persisting with white-nose syndrome.

Microbiome 5: 115.

Li, H., B. Handsaker, A. Wysoker, T. Fennell, J. Ruan *et al.*, 2009 The sequence

alignment/map format and SAMtools. Bioinforma. Oxf. Engl. 25: 2078–2079.

Lilley, T., J. Anttila, and L. Ruokolainen, 2018 Landscape structure and ecology influence

the spread of a bat fungal disease. Funct. Ecol. 32: 2483–2496.

Lilley, T. M., J. S. Johnson, L. Ruokolainen, E. J. Rogers, C. Wilson *et al.*, 2016 White-nose

syndrome survivors do not exhibit frequent arousals associated with

Pseudogymnoascus destructans infection. Front. Zool. 13: 12.

Lilley, T. M., J. M. Prokkola, A. S. Blomberg, S. Paterson, J. S. Johnson *et al.*, 2019

Resistance is futile: RNA-sequencing reveals differing responses to bat fungal

pathogen in Nearctic *Myotis lucifugus* and Palearctic *Myotis myotis*. Oecologia 295–

309.

Lilley, T. M., J. M. Prokkola, J. S. Johnson, E. J. Rogers, S. Gronsky *et al.*, 2017 Immune

responses in hibernating little brown myotis (*Myotis lucifugus*) with white-nose

syndrome. Proc R Soc B 284: 20162232.

Lips, K. R., F. Brem, R. Brenes, J. D. Reeve, R. A. Alford *et al.*, 2006 Emerging infectious

disease and the loss of biodiversity in a Neotropical amphibian community. Proc.

Natl. Acad. Sci. U. S. A. 103: 3165–3170.

Lorch, J. M., C. U. Meteyer, M. J. Behr, J. G. Boyles, P. M. Cryan *et al.*, 2011 Experimental

infection of bats with *Geomyces destructans* causes white-nose syndrome. Nature

480: 376-U129.

Lorch, J. M., L. K. Muller, R. E. Russell, M. O'Connor, D. L. Lindner *et al.*, 2013

Distribution and environmental persistence of the causative agent of white-nose

syndrome, *Geomyces destructans*, in bat hibernacula of the Eastern United States.

Appl. Environ. Microbiol. 79: 1293–1301.

Luikart, G., F. W. Allendorf, J.-M. Cornuet, and W. B. Sherwin, 1998 Distortion of allele

frequency distributions provides a test for recent population bottlenecks. J. Hered. 89:

238–247.

Luis, A. D., and P. J. Hudson, 2006 Hibernation patterns in mammals: a role for bacterial

growth? Funct. Ecol. 20: 471–477.

Marroquin, C. M., J. O. Lavine, and S. T. Windstam, 2017 Effect of humidity on

development of *Pseudogymnoascus destructans*, the causal agent of bat white-nose

syndrome. Northeast. Nat. 24: 54–64.

Martel, A., M. Blooi, C. Adriaensen, P. V. Rooij, W. Beukema *et al.*, 2014 Recent

introduction of a chytrid fungus endangers Western Palearctic salamanders. Science

346: 630–631.

Martin, M., 2011 Cutadapt removes adapter sequences from high-throughput sequencing
reads. EMBnet.journal 17: 10.

Maslo, B., and N. H. Fefferman, 2015 A case study of bats and white-nose syndrome
demonstrating how to model population viability with evolutionary effects. Conserv.
Biol. 29: 1176–1185.

Maslo, B., M. Valent, J. F. Gumbs, and W. F. Frick, 2015 Conservation implications of
ameliorating survival of little brown bats with white-nose syndrome. Ecol. Appl. 25:
1832–1840.

McGuire, L. P., H. W. Mayberry, and C. K. R. Willis, 2017 White-nose syndrome increases
torpid metabolic rate and evaporative water loss in hibernating bats. Am. J. Physiol.-
Regul. Integr. Comp. Physiol. 313: R680–R686.

McKenna, A., M. Hanna, E. Banks, A. Sivachenko, K. Cibulskis *et al.*, 2010 The Genome
Analysis Toolkit: a MapReduce framework for analyzing next-generation DNA
sequencing data. Genome Res. 20: 1297–1303.

Messer, P. W., and D. A. Petrov, 2013 Population genomics of rapid adaptation by soft
selective sweeps. Trends Ecol. Evol. 28: 659–669.

Meteyer, C. U., D. Barber, and J. N. Mandl, 2012 Pathology in euthermic bats with white
nose syndrome suggests a natural manifestation of immune reconstitution
inflammatory syndrome. Virulence 3: 583–588.

Micalizzi, E. W., J. N. Mack, G. P. White, T. J. Avis, and M. L. Smith, 2017 Microbial
inhibitors of the fungus *Pseudogymnoascus destructans*, the causal agent of white-
nose syndrome in bats. PLOS ONE 12: e0179770.

Miller, C. T., S. Beleza, A. A. Pollen, D. Schluter, R. A. Kittles *et al.*, 2007 Cis-regulatory

changes in Kit ligand expression and parallel evolution of pigmentation in

sticklebacks and humans. Cell 131: 1179–1189.

Miller-Butterworth, C. M., M. J. Vonhof, J. Rosenstern, G. G. Turner, and A. L. Russell,

2014 Genetic structure of little brown bats (*Myotis lucifugus*) corresponds with spread

of white-nose syndrome among hibernacula. J. Hered. 105: 354–364.

Minnis, A. M., and D. L. Lindner, 2013 Phylogenetic evaluation of Geomyces and allies

reveals no close relatives of Pseudogymnoascus destructans, comb. nov., in bat

hibernacula of eastern North America. Fungal Biol. 117: 638–649.

Morales, A. E., and B. C. Carstens, 2018 Evidence that *Myotis lucifugus* "subspecies" are

five nonsister species, despite gene flow. Syst. Biol. 67: 756–769.

Norquay, K. J. O., F. Martinez-Nuñez, J. E. Dubois, K. M. Monson, and C. K. R. Willis,

2013 Long-distance movements of little brown bats (*Myotis lucifugus*). J. Mammal.

94: 506–515.

Olivera-Hyde, M., A. Silvis, E. M. Hallerman, W. M. Ford, and E. R. Britzke, 2019

Relatedness within and among northern long-eared bat (*Myotis septentrionalis*)

colonies at a local scale. Can. J. Zool. 97: 724–735.

Paterson, S., T. Vogwill, A. Buckling, R. Benmayor, A. J. Spiers *et al.*, 2010 Antagonistic

coevolution accelerates molecular evolution. Nature 464: 275–278.

Paterson, S., K. Wilson, and J. M. Pemberton, 1998 Major histocompatibility complex

variation associated with juvenile survival and parasite resistance in a large

744  unmanaged ungulate population (*Ovis aries* L.). Proc. Natl. Acad. Sci. 95: 3714–

745  3719.

746 Peters, J.-M., A. Tedeschi, and J. Schmitz, 2008 The cohesin complex and its roles in

747  chromosome biology. Genes Dev. 22: 3089–3114.

748 Puechmaille, S. J., G. Wibbelt, V. Korn, H. Fuller, F. Forget *et al.*, 2011 Pan-European

749  distribution of white-nose syndrome fungus (*Geomyces destructans*) not associated

750  with mass mortality. PLoS ONE 6: e19167.

751 R Development Core Team, 2011 R: A language and environment for statistical computing.

752 Råberg, L., A. L. Graham, and A. F. Read, 2009 Decomposing health: tolerance and

753  resistance to parasites in animals. Philos. Trans. R. Soc. Lond. B. Biol. Sci. 364: 37–

754  49.

755 Reeder, D. M., C. L. Frank, G. G. Turner, C. U. Meteyer, A. Kurta *et al.*, 2012 Frequent

756  arousal from hibernation linked to severity of infection and mortality in bats with

757  white-nose syndrome. Plos One 7: e38920.

758 Reeder, S. M., J. M. Palmer, J. M. Prokkola, T. M. Lilley, D. M. Reeder *et al.*, 2017

759  Pseudogymnoascus destructans transcriptome changes during white-nose syndrome

760  infections. Virulence 8: 1695–1707.

761 Ren, P., K. H. Haman, L. A. Last, S. S. Rajkumar, M. K. Keel *et al.*, 2012 Clonal spread of

762  *Geomyces destructans* among bats, Midwestern and Southern United States. Emerg.

763  Infect. Dis. 18: 883–885.

Rico, Y., J. Morris-Pocock, J. Zigouris, J. J. Nocera, and C. J. Kyle, 2015 Lack of spatial

immunogenetic structure among wolverine (*Gulo gulo*) populations suggestive of

broad scale balancing selection. PLOS ONE 10: e0140170.

van Riper, C., S. G. van Riper, M. L. Goff, and M. Laird, 1986 The epizootiology and

ecological significance of malaria in Hawaiian land birds. Ecol. Monogr. 56: 327–

344.

Robinson, S. J., M. D. Samuel, C. J. Johnson, M. Adams, and D. I. McKenzie, 2012

Emerging prion disease drives host selection in a wildlife population. Ecol. Appl.

Publ. Ecol. Soc. Am. 22: 1050–1059.

Routtu, J., and D. Ebert, 2015 Genetic architecture of resistance in *Daphnia* hosts against two

species of host-specific parasites. Heredity 114: 241–248.

Schlötterer, C., R. Tobler, R. Kofler, and V. Nolte, 2014 Sequencing pools of individuals —

mining genome-wide polymorphism data without big funding. Nat. Rev. Genet. 15:

749–763.

Scholthof, K.-B. G., 2007 The disease triangle: pathogens, the environment and society. Nat.

Rev. Microbiol. 5: 152–156.

Schoville, S. D., A. Bonin, O. François, S. Lobreaux, C. Melodelima *et al.*, 2012 Adaptive

genetic variation on the landscape: methods and cases. Annu. Rev. Ecol. Evol. Syst.

43: 23–43.

Sironi, M., R. Cagliani, D. Forni, and M. Clerici, 2015 Evolutionary insights into host–

pathogen interactions from mammalian sequence data. Nat. Rev. Genet. 16: 224–236.

Smith, K. F., D. F. Sax, and K. D. Lafferty, 2006 Evidence for the role of infectious disease

in species extinction and endangerment. Conserv. Biol. 20: 1349–1357.

Sparks, A. M., K. Watt, R. Sinclair, J. G. Pilkington, J. M. Pemberton *et al.*, 2019 The

genetic architecture of helminth-specific immune responses in a wild population of

Soay sheep (Ovis aries). bioRxiv 628271.

Spurgin, L. G., and D. S. Richardson, 2010 How pathogens drive genetic diversity: MHC,

mechanisms and misunderstandings. Proc. Biol. Sci. 277: 979–988.

Studier, E. H., and M. J. O'Farrell, 1972 Biology of *Myotis thysanodes* and *M. lucifugus*

(Chiroptera: Vespertilionidae). I. Thermoregulation. Comp. Biochem. Physiol. A 41:

567–595.

Teacher, A. G. F., T. W. J. Garner, and R. A. Nichols, 2009 Population genetic patterns

suggest a behavioural change in wild common frogs (*Rana temporaria*) following

disease outbreaks (Ranavirus). Mol. Ecol. 18: 3163–3172.

The UniProt Consortium, 2019 UniProt: a worldwide hub of protein knowledge. Nucleic

Acids Res. 47: D506–D515.

Thogmartin, W. E., R. A. King, P. C. McKann, J. A. Szymanski, and L. Pruitt, 2012

Population-level impact of white-nose syndrome on the endangered Indiana bat. J.

Mammal. 93: 1086–1098.

Thomas, D. W., M. Dorais, and J.-M. Bergeron, 1990 Winter energy budgets and cost of

arousals for hibernating little brown bats, *Myotis lucifugus*. J. Mammal. 71: 475–479.

Tompkins, D. M., S. Carver, M. E. Jones, M. Krkošek, and L. F. Skerratt, 2015 Emerging

infectious diseases of wildlife: a critical perspective. Trends Parasitol. 31: 149–159.

Turner, G. G., D. M. Reeder, and J. T. H. Coleman, 2011 A five-year assessment of mortality

and geographic spread of white-nose syndrome in North American bats and a look to

the future. Bat Res. News 52: 13–27.

Verant, M. L., J. G. Boyles, W. Waldrep, G. Wibbelt, and D. S. Blehert, 2012 Temperature-

dependent growth of *Geomyces destructans*, the fungus that causes bat white-nose

syndrome. Plos One 7: e46280.

Verant, M. L., C. U. Meteyer, J. R. Speakman, P. M. Cryan, J. M. Lorch *et al.*, 2014 White-

nose syndrome initiates a cascade of physiologic disturbances in the hibernating bat

host. BMC Physiol. 14: 10.

Vonhof, M. J., A. L. Russell, and C. M. Miller-Butterworth, 2015 Range-wide genetic

analysis of little brown bat (*Myotis lucifugus*) populations: estimating the risk of

spread of White-nose syndrome. PLoS ONE 10: e0128713.

Warnecke, L., J. M. Turner, T. K. Bollinger, J. M. Lorch, V. Misra *et al.*, 2012 Inoculation of

bats with European *Geomyces destructans* supports the novel pathogen hypothesis for

the origin of white-nose syndrome. Proc. Natl. Acad. Sci. U. S. A. 109: 6999–7003.

Warnecke, L., J. M. Turner, T. K. Bollinger, V. Misra, P. M. Cryan *et al.*, 2013

Pathophysiology of white-nose syndrome in bats: a mechanistic model linking wing

damage to mortality. Biol. Lett. 9: 20130177.

Warnecke, L., J. M. Turner, and F. Geiser, 2008 Torpor and basking in a small arid zone

marsupial. Naturwissenschaften 95: 73–78.

Wibbelt, G., A. Kurth, D. Hellmann, M. Weishaar, A. Barlow *et al.*, 2010 White-nose

syndrome fungus (*Geomyces destructans*) in Bats, Europe. Emerg. Infect. Dis. 16:

1237–1243.

Wibbelt, G., S. J. Puechmaille, B. Ohlendorf, K. Muehldorfer, T. Bosch *et al.*, 2013 Skin

lesions in European hibernating bats associated with *Geomyces destructans*, the

etiologic agent of white-nose syndrome. Plos One 8: e74105.

Wilder, A. P., T. H. Kunz, and M. D. Sorenson, 2015 Population genetic structure of a

common host predicts the spread of white-nose syndrome, an emerging infectious

disease in bats. Mol. Ecol. 24: 5495–5506.

Willis, C. K. R., A. K. Menzies, J. G. Boyles, and M. S. Wojciechowski, 2011 Evaporative

water loss is a plausible explanation for mortality of bats from white-nose syndrome.

Integr. Comp. Biol. 51: 364–373.

Wright, S., 1978 *Evolution and the Genetics of Populations, Volume 4. Variability within and*

among Natural Populations. University of Chicago Press, Chicago.

www.whitenosesyndrome.org, 2020 White-nose syndrome occurance map - by year (2019).

Data   last   updated: 3/27/2020.   Available   at:

https://www.whitenosesyndrome.org/resources/map.

Yang, Z., H. Shi, P. Ma, S. Zhao, Q. Kong *et al.*, 2018 Darwinian positive selection on the

pleiotropic effects of KITLG explain skin pigmentation and winter temperature

adaptation in Eurasians. Mol. Biol. Evol. 1–12.

Zeisset, I., and T. J. C. Beebee, 2014 Drift rather than selection dominates MHC Class II

allelic diversity patterns at the biogeographical range scale in natterjack toads *Bufo*

calamita. PLOS ONE 9: e100176.

Zhang, T., V. Chaturvedi, and S. Chaturvedi, 2015 Novel *Trichoderma polysporum* strain for

the biocontrol of *Pseudogymnoascus destructans*, the fungal etiologic agent of bat

white nose syndrome. PLOS ONE 10: e0141316.

Zhang, T., T. R. Victor, S. S. Rajkumar, X. Li, J. C. Okoniewski *et al.*, 2014 Mycobiome of

the bat white nose syndrome affected caves and mines reveals diversity of fungi and

local adaptation by the fungal pathogen *Pseudogymnoascus (Geomyces) destructans*.

PLOS ONE 9: e108714.

Zukal, J., H. Bandouchova, J. Brichta, A. Cmokova, K. S. Jaron *et al.*, 2016 White-nose

syndrome without borders: *Pseudogymnoascus destructans* infection tolerated in

Europe and Palearctic Asia but not in North America. Sci. Rep. 6: 19829.

FIGURE LEGENDS

Fig. 1. *Myotis lucifugus* sampling sites in Pennsylvania (PA) and New York (NY). Pre-WNS-

sites in grey circles and Post-WNS-sites in white circles. Point 5 is Hailes Cave, near the first

point of discovery of WNS. Site numbers correspond to Supplemental Table 1. Sample

numbers per site 1) n=3; 2) n=19; 3) n=3; 4) n=2; 5) n=24; 6) n=32; 7) n=14; 8) n=20; 9)

n=4; 10) n=12; 11) n=12; 12) n=16; 13) n=8; 14) n=36; and 15) n=21.

Fig. 2. Fixation indices (F_{ST}) across the *M. lucifugus* genome, comparing the population

before and after the arrival of white-nose syndrome. Scaffold lengths presented in green and

blue. Solid red line indicates cut-off for F_{ST} values of 5 standard deviations from the mean

across the whole genome.

Fig 3. Fixation indices (F_{ST}) across the *M. lucifugus* genome between geographically

separated individuals of the study population: A) comparing individuals from PA to

individuals from NY before the arrival of WNS; B) comparing individuals from PA before

and after the arrival of WNS; and C) comparing individuals from NY before and after the

arrival of WNS. Scaffold lengths presented in green and blue. Solid red line indicates cut-off

for F_{ST} values of 5 standard deviations from the mean across the whole genome; dotted red

line indicates cut-off of F_{ST}=0.05. Windows exceeding this cutoff of F_{ST}=0.05 are colored

red.

Fig 4. Fixation indices (F_{ST}) across the *M. lucifugus* genome between geographically

separated individuals of the study population: A) comparing individuals from PA to

individuals from NY after the arrival of WNS, with a zoomed plot of scaffold GL429776 and

gene models therein containing windows demonstrating F_{ST} values > 0.05; and B) comparing

individuals from PA to individuals from NY with Pre- and Post-WNS data combined.

Scaffold lengths presented in green and blue. Solid red line indicates cut-off for F_{ST} values of

5 standard deviations from the mean across the whole genome; dotted red line indicates cut-

off of F_{ST}=0.05. Windows exceeding this cutoff of F_{ST}=0.05 are colored red.

Fig 5. Proportions of reference alleles called at individual SNP sites for the region of scaffold

GL429776 demonstrating high F_{ST} values when comparing: A) individuals from

Pennsylvania before and after the arrival of WNS; B) individuals from NY before and after

the arrival of WNS; C) individuals from PA to individuals from NY before the arrival of

WNS; and D) individuals from PA to individuals from NY after the arrival of WNS.

Table 1. Samples used in study.

Sample age	Classification	Total number of samples	Pennsylvania*	New York*
2006	Pre-WNS	25	25	
2007	Pre-WNS	27	4	24
2015	Post-WNS	28	28	
2016	Post-WNS	139	82	57

* WNS first detected in New York in 2006 and PA in 2009

Table 2. Summary of individual SNP F_{ST} values across the *M. lucifugus* genome when comparing Pre-WNS with Post-WNS; and samples from PA and NY (both Pre- and Post-WNS).

	Pre- vs post-WNS	Between site (combining pre- and post-WNS)
Median (Interquartile range)	0.0059 (0.0013 – 0.0171)	0.0057 (0.0013 – 0.0165)
Mean (s.d.)	0.0127 (0.0173)	0.0123 (0.0170)

Table 3. Genes within the *M. lucifugus* genome – excluding scaffold GL429776 – with windows possessing F_{ST} values >= 0.05 when comparing PA and NY samples Post-WNS.

Scaffold	Location range (bp)	Gene ID	Protein Function	Number of overlapping high F_{ST} windows
NW_005871715.1	288334 - 351644	102435296	bifunctional heparan sulfate N-deacetylase/N-sulfotransferase 4-like (partial)	4
NW_005871771.1	26662 - 48938	102441697	E3 ubiquitin-protein ligase TRIM7	1
NW_005871823.1	81353 - 213965	102428352	Ankyrin 2	7
NW_005871905.1	175 - 139344	102428043	calcium/calmodulin-dependent protein kinase	10

			type II subunit delta-like	
NW_005872026.1	72042 - 101269	102443179	N-deacetylase and N-sulfotransferase 4 (partial)	1
NW_005872062.1	33266 - 86607	102428048	zinc finger GRF-type containing 1	1

Table 4. Genes within scaffold GL429776 with windows possessing F_{ST} values >= 0.05 when comparing PA and NY samples Post-WNS.

Gene ID	F_{ST}	No of windows	No of windows (per 10kb)	Start pos.	End pos.	Details
MGAT4C	0.0612	78	1.08	8999210	9724150	Glycosyltransferase that participates in the transfer of N-acetylglucosamine (GlcNAc) to the core mannose residues of N-linked glycans
CEP290	0.0679	12	1.01	10709418	10827912	Involved in early and late steps in cilia formation
ATP2B1	0.0826	12	0.89	12228178	12362389	The protein encoded by this gene belongs to the family of P-type primary ion transport ATPases characterised by the formation of an aspartyl phosphate intermediate during the reaction cycle. These enzymes remove bivalent

						calcium ions from eukaryotic cells against very large concentration gradients and play a critical role in intracellular calcium homeostasis.
KITLG	0.0706	10	1.13	11127130	11215731	Plays an essential role in the regulation of cell survival and proliferation, hematopoiesis, stem cell maintenance, gametogenesis, mast cell development, migration and function, and in melanogenesis.
POC1B	0.0686	7	0.74	12044678	12139811	Plays an important role in centriole assembly and/or stability and ciliogenesis
TMTC3	0.0872	6	1.00	10827719	10887502	Involved in the positive regulation of proteasomal protein degradation in the endoplasmic reticulum (ER), and the control of ER stress response.
GeneID:111824954	0.0693	5	1.14	11756928	11800690	uncharacterised LOC111824954

CUNH12orf50	0.0588	2	0.67	10663328	10693373	chromosome unknown C12orf50 homolog
GeneID:111824948	0.0816	2	1.13	11375011	11392720	uncharacterised LOC111824948
GeneID:111824949	0.0635	2	1.14	11802704	11820273	uncharacterised LOC111824949
CUNH12orf29	0.0488	1	0.71	10696290	10710421	chromosome unknown C12orf29 homolog
DUSP6	0.0603	1	2.26	11976039	11980455	Inactivates MAP kinases.
GeneID:111824955	0.0459	1	1.57	12101754	12108133	uncharacterised LOC111824955