ORIGINAL PAPER



Bucking the trend of pollinator decline: the population genetics of a range expanding bumblebee

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Received: 1 July 2020 / Accepted: 2 March 2021 © The Author(s), under exclusive licence to Springer Nature Switzerland AG 2021

Abstract

Recent research has shown drastic reductions in the global diversity and abundance of insects. This is a major concern given the expected cascade effects on ecosystem services, such as pollination. Understanding the patterns and drivers of changes in the distribution and abundance of species in our rapidly changing environment is therefore urgent. Cases of species showing trends that run counter to general population declines, especially when they deliver key ecosystem services, are especially interesting. The tree bumblebee (Bombus hypnorum), which belongs to a globally important group of pollinators, has substantially expanded its range in recent years in direct contrast to many other species within this group. Here we reconstructed the likely pattern of colonization of the UK based on RADseq population genomic data combined with Bayesian population modelling. This RADseq approach also enabled an analysis of genomic regions potentially under selection. We report a complex and dynamic colonization pattern that is most likely ongoing. Current evidence suggests that either a shift in its migration potential, and/or adaptive genomic changes have contributed to the recent range expansion of B. hypnorum. Genomic areas of potential adaptive significance included genes involved in regulation of transcription and gene expression, circadian rhythms and innate immunity. Our results are framed within the general context of understanding the factors driving successful population expansions.

Keywords Bombus hypnorum \cdot Range expansion \cdot Pollinators \cdot Invasion genetics \cdot ABC population modelling \cdot RAD-seq

Introduction

Current rates of species extinction are substantially elevated relative to the historical record (Ceballos et al. 2015) and many species are undergoing declines in range and abundance (Hallmann et al. 2017). However, some species remain widespread or are even undergoing

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range expansion. Such contrasting patterns of population decline and success are especially interesting when they occur among species within close phylogenetic groups (McKinney and Lockwood 1999; Angert et al. 2011; Moran and Alexander 2014). This allows identification of traits that potentially contribute to species' resilience or extinction risk (Purvis 2008; Chichorro et al. 2019) and therefore provides ideal opportunity to gain traction on the factors driving diversity changes (Sax et al. 2007; Moran and Alexander 2014).

A key aspect of understanding these rapid changes in global biodiversity patterns is unravelling the dynamics of colonization. This has historically been achievable only through reconstructing events from observational records. While useful, these are not always available and are typically limited in their capacity to trace historical events with accuracy. However, the progressive accessibility of genome-scale data, and development of population modelling tools (Cornuet et al. 2014; Cabrera and Palsbøll 2017), have offered a powerful approach that is yielding significant insights into the routes and biological signatures of colonizations (e.g. Guzinski et al. 2018). These tools have been successfully used in a range of contexts, such as inference of the evolutionary history of speciation (Momigliano et al. 2017), measuring the success of translocation programmes (Puckett et al. 2014), and investigating meta-population dynamics (Stillfried et al. 2017).

Advances in sequencing technology also facilitate the investigation of the wider genetic effects associated with range expansions. For example, new colonizations are often characterized by a limited number of founders, resulting in a population bottleneck and small initial population sizes (Dlugosch and Parker 2008). However, colonizing species often successfully establish within their new environment despite the predicted loss of genetic diversity and harmful inbreeding effects, posing a genetic paradox (Allendorf and Lundquist 2003; Sax and Brown 2000; Estoup et al. 2016). One potential explanation is that bottlenecks may not lead to a substantial loss of genetic variance in quantitative traits (Lewontin 1965); it may even increase when dominance interactions are considered (Robertson 1952). The loss of genetic diversity during colonization may also be overcome if there are multiple colonization events or high migration rates from the source population, which can eliminate founder effects (Lockwood et al. 2005; Roman and Darling 2007; Dlugosch and Parker 2008).

In parallel with developments in sequencing technologies, there have been theoretical advances in the understanding of range expansions. Facon et al. (2006) proposed a useful framework to consider colonization scenarios, identifying three scenarios that are not mutually exclusive. In the first, 'migration change', a species is limited only by its capacity to migrate to an existing habitat that meets its requirements. Here, colonization may occur if barriers to migration are removed through, for example, human activities. The second scenario, 'environmental change', describes the situation where suitable habitat becomes newly available. The third scenario, 'evolutionary change', assumes genetic changes in the colonizing species that either occur prior to, or during, the colonization, conferring a fitness advantage in the new environment. Successful range expansions following evolutionary change have been linked to adaptations to anthropogenically modified habitats, coined as 'Anthropogenically Induced Adaptations to Invade' (Hufbauer et al. 2012).

Bumblebees (*Bombus* spp.) are a highly pertinent group to investigate factors driving changes in diversity as the group contains taxa with strikingly different population trends. In addition, globally, patterns of decline are predominant, which is an issue of particular concern because of their role as ecologically and economically important pollinators (Goulson et al. 2005; Williams 2005; Williams and Osborne 2009).

Patterns of declines with the *Bombus* group are, at least to some degree, phylogenetically structured. For example, members of the subgenus *Thoracobombus* appear to show

increased vulnerability to population decline (Cameron et al. 2011; Arbetman et al. 2017), as do several others (Goulson et al. 2008), but those of the subgenus *Pyrobombus*, in contrast, seem to exhibit increased resilience (Arbetman et al. 2017; Richardson et al. 2019). Several *Pyrobombus* species are reported to be increasing in abundance and/or expanding their range, for example *Bombus haematurus* in central Europe (Biella 2020), *Bombus pratorum* and *Bombus monticola* which have colonized Ireland during the last century (Speight 1974; Fitzpatrick et al. 2007), *Bombus bimaculatus, Bombus impatiens, Bombus ternarius*, and *Bombus vagans* in Vermont (Richardson et al. 2015). These contrasting patterns suggest that this group exhibits traits that make it less vulnerable to threats that are causing declines in the majority of other bumblebees. Investigating the underlying mechanisms for this increased resilience in this group are therefore important to improve understanding of the drivers of diversity change, and in turn inform conservation efforts, in this important group of pollinators.

A notably successful species among the *Pyrobombus* group is the tree bumblebee, *Bom*bus hypnorum, which has recently substantially expanded its range (Goulson and Williams 2001; Prŷs-Jones et al. 2016). Bombus hypnorum is one of the most widespread bumblebee species across Europe and Asia (Williams 1991; Goulson and Williams 2001), with a broad palaearctic distribution from Iceland to Japan (Williams 1991; Rasmont et al. 2015). It has been present along the north-western coast of Belgium and France from at least the early twentieth century (Rasmont 1988). A recent study showed a lack of genetic structuring of B. hypnorum populations in Belgium, indicating a large panmictic population across western parts of Europe (Maebe et al. 2019). Significant differentiation was observed between western European and Baltic populations, suggesting population structure at much larger geographical scales (Maebe et al. 2019). In the last decades, B. hypnorum has expanded its range significantly in the western part of its distribution (Rasmont et al. 2015), reaching Iceland in 2010 (Prŷs-Jones et al. 2016) and Ireland in 2017. It was first recorded in the south of England in 2001 (Goulson and Williams 2001) and has since spread rapidly northwards, arriving in Scotland in 2012. It is now one of the most common bumblebee species in the UK, both abundant and widely distributed (BWARS 2019). Given no published data to-date on any negative impacts of this population expansion, we refer to this throughout as a colonization rather than an invasion.

There are a number of characteristics of B. hypnorum that may indicate its potential as a successful colonizer. It has a wide distribution across Europe and Asia, where it is found across diverse types of habitats (Goulson and Williams 2001). This reflects its broad niche, including dietary and climatic requirements (BWARS 2019), attributes generally associated with successfully colonizing species (Baker 1965; Willamson and Fitter 1996; Vazquez 2006). It exhibits a facultative bivoltine colony cycle (Edwards and Jenner 2005), which may contribute to an increased rate of population growth facilitating an accelerated spread (Sakai et al. 2001). It also shows facultative polyandrous mating patterns, at least in some parts of it range, which may increase the genetic diversity within colonies (Paxton et al. 2001). Finally, B. hypnorum is unique amongst UK Bombus species in nesting high above ground (Benton 2006), often in buildings, bird nest boxes and roof structures, which may give it a competitive advantage in a highly urbanized environment (Crowther et al. 2014). However, what has enabled its rapid spread since 2001 remains unclear: these are longstanding characteristics and B. hypnorum has nonetheless only very recently colonized the UK. Information on continental populations is incomplete, but available information points to B. hypnorum being well established and abundant across western continental Europe for some time (>100 years) prior to its colonization of the UK (Rasmont et al. 1988; Maebe et al. 2019). Although there are intrinsic lag times in the introduction, growth, expansion and detection of colonizing species (Crooks et al. 2005), this lag would be particularly prolonged (up to 200 generation since *B. hypnorum* is bivoltine) in this case if the colonization of the UK represents the continuation of a range expansion dating back to over a century ago. Thus, evidence to-date points to a significant change in either the environment, which includes an increase in opportunities for migration, e.g. through human aided transport, or the intrinsic biology of this species that have contributed to its sudden and highly successful range expansion.

Here, we explore the genetic signatures associated with this rapid expansion of *B. hypnorum* into the UK using a RAD-seq population genomic dataset. Our specific goals were to: (i) identify whether the UK population was founded from a single event or from multiple and potentially ongoing events; (ii) assess population structure and whether there has been any significant loss of genetic diversity as a consequence of the population expansion; (iii) identify preliminary indicators of any signatures of selection before or during the colonization that may have promoted its success.

Materials and methods

Establishing the spread of Bombus hypnorum across the UK

The likely geographical spread of *Bombus hypnorum* across the UK from its first record in 2001 (Goulson and Williams 2001) was reconstructed using the BWARS database (Bees, Wasps and Ants Recording Society 2019), which collates recorded sightings as part of an ongoing dedicated mapping program (Fig. 1). This database is reliant on records being sent in by the public, amateur groups and specialists. Although it will inevitably contain some gaps and biases (e.g. towards human population density), all records are verified by specialists within BWARS for quality assurance (van der Wal et al. 2015). While not systematic, it is nevertheless likely to estimate broad patterns reasonably, and has been used previously for such purposes (e.g. Potts et al. 2010; Ollerton et al. 2014).

Tissue samples and RAD library preparation

Bombus hypnorum was sampled from seven localities representing different 'fronts' of the range expansion of this species (as established from the BWARS database, see also Results). Six localities were sampled in the UK (Cardiff, Hull, London, Newcastle, Plymouth and Southampton), representing different stages of the expansion, and one in France (Le Havre) representing its already well-established distribution in continental Europe (Maebe et al. 2019).

Forty individual samples were collected across a large area (approximately 10×10 km) at each site using standard population sampling methods for colony-living Hymenopteran species (see e.g. Goulson et al. 2011) where individual samples were collected a minimum of 200 m apart to avoid any significant sampling of sisters (belonging to the same nest). Samples were collected between May and July of 2013 and 2014 and stored in 100% ethanol.

DNA was extracted from the thoracic muscle tissue of all sampled individuals using an ammonium acetate protocol (Nicholls et al. 2000) and quantified on a Qubit 3.0



Fig. 1 Reconstruction of the UK Bombus hypnorum colonization from BWARS records of year-wise sightings 2004–2012

Fluorometer using a broad range assay (Thermo Fisher Scientific). DNA (500 ng) was digested in 20 µl volumes after RNase treatment with 40 units of the restriction enzyme Xhol (New England Biolabs) at 37 °C for 3 h with a 20 min heat deactivation stage at 80 °C. Digested DNA was purified using AmpureXP (Beckmann and Coulter, 1.4X ratio of beads to DNA) and quantified on a Qubit 3.0 Fluorometer (Thermo Fisher Scientific). RAD

libraries were prepared using the IonXpress Plus gDNA Fragment Library Kit (Life Technologies). Library preparation was carried out using a pooled approach with populationspecific barcodes using equimolar concentrations from each individually digested sample for each sampling site. A pooled sequencing design represents a well established and cost-effective alternative to individual sequencing to reliably obtain genome wide allele frequency data (Futschik and Schlötterer 2010; Gautier et al. 2013; Schlötterer et al. 2014) and has been used in a wide range of systems (e.g. Guo et al. 2015; Kahnt et al. 2018). (We note that a limitation of this approach is the lack of individual genotypes and heterozygosity necessary for analysis such as assignment tests or estimation of inbreeding coefficients (Andrews et al. 2016)). For quality control, library preparation of the Newcastle sample was performed twice independently using the pooled approach. Additionally, both the Newcastle and Le Havre samples were prepared and sequenced using an individual barcoding approach with twenty individual samples for each sampling site (Gautier et al. 2013). A barcoded Ion Torrent adapter A was ligated using 0.1 µM of barcode adaptor, 200 U of T4 DNA ligase (New England Biolabs), 100 mM of ATP and 2 µl of NE 4 Buffer in 40 µl volumes for 2 h at 22 °C, followed by heat deactivation at 65 °C for 20 min, either on an individual or pooled basis for each sampling site. Purification was repeated twice after this step (1.2X ratio of beads to DNA) and prior to shearing using Ion Shear Plus Enzyme Mix II (Life Technologies) following the manufacturer's protocol. After further AmpureXP purification (1.4X ratio of beads to DNA) the Ion Torrent adapter P1 (Thermo Fisher Scientific) was ligated in 49 µl volumes for 20 min at 25 °C followed by heat deactivation at 72 °C for 5 min following the manufacturer's guidelines. After another step of AmpureXP purification (1.2X ratio of beads to DNA), library amplification was achieved through: 5 min at 95 °C, followed by 18 cycles of a heat denaturation at 95 °C for 15 s, annealing at 58 °C for 15 s and extension at 70 °C for 1 min. Pippin Prep (Sage Science) was employed to select for a fragment size range of 159 and 164 bp prior to AmpureXP purification (1.5X ratio of beads to DNA) and quantification was then estimated by qPCR. Libraries were run on an Ion Torrent PGM using one 318 chip for each individually barcoded sample or for two pooled samples respectively.

Data processing and SNP calling

Raw reads were trimmed from both ends to excise low quality base-calls (average Q-score < 15 across 4 bp sliding windows) and filtered for a minimum length of 10 bp using Trimmomatic-0.36 (Bolger et al. 2014). Filtered reads were aligned to the Bombus terrestris genome (Bter_1.0 assembly, Ensembl) using the Burrow-Wheeler Aligner (BWA) and the MEM algorithm (Li 2013), which has been shown to perform best in the presence of indel errors typical of the Ion Torrent (Ziemann 2016). We chose the Bombus terrestris genome, and not that of the more closely related *Bombus impatiens*, because linkage group annotations are available for this species. This allowed us to investigate patterns of diversity across genomic regions. We tested if there was evidence for bias caused by this choice through comparison of mapping statistics against the *Bombus impatiens* genome. Aligned reads were only retained if they mapped uniquely to the reference genome and had a minimum mapping quality score of 20 (Cibulskis et al. 2013). Aligned BAM files were sorted and converted into the mpileup format allowing a maximum of 1,000 reads at a given position using Samtools 0.1.5 (Li et al. 2009). SNP calling (excluding indels) for individually barcoded samples was performed using VarScan v.2.3.9 (Koboldt et al. 2009) using all sites that had a minimum coverage of 10 in at least 10 individuals. For these individual samples, requirements for SNP acceptance were a minimum of two reads (default) for the minor allele and a p-value of ≤ 0.05 derived from Fisher's exact test on read counts (coverage) supporting the reference and variant allele respectively (Koboldt et al. 2009, 2013).

The mpileup file of pooled samples was converted to the sync file format and filtered for indels using Popoolation2 (Kofler et al. 2011). Because variation in sequencing depth can impair the accuracy of allele frequency estimates derived from pooled samples, only sites with a sequencing coverage of ≥ 50 and ≤ 500 within any sample, and a maximum coverage of 3000 across all samples, were considered for pooled samples. Note that specifically for the analysis of the distribution of genome wide polymorphism, which does not rely on allele frequency estimates, a minimum within-sample coverage of 10 was applied. For pooled samples, SNPs were called if they had a minimum of 3 reads per allele across all samples and a minimum within-sample allele frequency of 0.01, as commonly used in population genetic studies (e.g. Bruneaux et al. 2013; Fraser et al. 2014). We also applied a threshold of 0.05 for the identification of F_{ST} -outliers to test for consistency (Roesti et al. 2012).

Analysis of population genetic diversity and characterization of population structure

Average expected genome-wide heterozygosity was calculated following Fischer et al. (2017). The proportion of polymorphic sites was calculated across all base pairs covered in all samples. Pseudo-haplotypes were created based on population allele frequencies using a custom Python script and imported into R v.3.4.2 (R Core Team, 2012). Allelic richness was calculated in the R package Hierfstat v.0.04-22 (Goudet 2004). Pairwise population $F_{\rm ST}$ was calculated using Popoolation2 (Kofler et al. 2011) and confidence intervals were created by bootstrapping over loci for 1000 cycles using a custom Python script. A Mann–Whitney-Wilcoxon test was used to test if there is a significant difference in the pairwise $F_{\rm ST}$ values between UK-Le Havre and UK-UK populations (Mann and Whitney 1947). A Mantel test (Mantel 1967) was implemented in Genepop 4.2 (Raymond and Rousset 1995) to test for correlation between geographic and genetic distance. A principal coordinate analysis was performed in the R package Hierfstat v.0.04-22 (Goudet 2004). This was carried out for SNPs covered in all sample sites, a minimum of six sample sites (75%) and a minimum of four sample sites (50%) to assess the effect of missing data on the resolution of population structure.

Testing hypotheses of colonization history

The approximate Bayesian computation software DIYABC v2.0 (Cornuet et al. 2014) was used to assess the relative probability of different colonization scenarios. These were derived using a 'bottom up' hierarchical nested approach to identify the most likely colonization for each UK sample site independently (described in detail below). The highest probability scenario for each individual sampling site was combined into a final consensus model. The advantages of this approach are that it allows the most robust model to be constructed from the data available. No prior assumptions based on the observed BWARS records were required and the history of each sampling site is built independently. It also allowed model construction using the maximum number of available SNPs at each stage. This approach involved the following steps:

Step 1 For each UK sampling site separately, the likelihood of the panmictic Western European population (sampled at Le Havre, France) as a colonization source was tested in DIYABC against a simulated population, where the Western European population and the simulated population share the same common ancestral population (Fig. 2). This was carried out to test for the contribution of a potential other source that is more divergent to the Western European population, e.g. from Scandinavia. The possibility of simulating unsampled populations (with shared ancestry to test populations) is an effective feature of the DIYABC software implementation to account for the possible contribution of other genetic lineages that have not been sampled (Estoup and Guillemaud 2010). Simulated populations are constructed using coalescence theory by generating a genealogy based on the defined order of events in a given scenario (Cornuet et al. 2014).

Step 2 This next step tested whether each UK sampling site was (i) founded independently from the most likely source identified in step 1; (ii) was founded sequentially via the most likely source in step one and through another UK sampling site (e.g. Le Havre colonized Southampton which colonized London); (iii) was founded by ongoing colonization from the source identified in step one and concurrent colonization by another UK sampling site (Fig. 2).

Step 3 In the case that more than one UK sampling site was more likely to be the source of another UK sample than the continental sample in step 2, these were tested against each other as well as against a dual (concurrent) colonization from each of them. As an example: for the Cardiff sampling site, the most likely scenario derived from step 1 was a colonization from Le Havre. Testing this scenario against the probability of colonization from every other UK sampling site directly or in combination with Le Havre resulted in both Hull and Newcastle being more likely as a source of colonization than Le Havre. In step 3 a colonization from Hull was tested against a colonization from Newcastle and a dual colonization from both sites with the latter showing highest probability.

Step 4 The most likely scenarios from steps 1–3 were then combined into a final consensus model that included all sample sites.

All scenarios implemented were tested with or without a bottleneck.

Each individual scenario was tested using a total of 100,000 iterations, following recommended guidelines (Cornuet et al. 2014). For Newcastle (which included a technical



Fig.2 schematic representation of steps 1 and 2 of the hierarchical DIYABC analysis done for each UK sampling site independently

replicate), only the "Newcastle 2" replicate was considered (see "Results"). Model evaluation was based on summary statistics for all biallelic SNP markers covered in the sampling sites considered. Summary statistics consisted of the mean of non-zero values, variance of non-zero values and mean of the complete distribution for genic diversity and pairwise F_{ST} and Nei's distance. Uniform simulation priors were applied to all demographic parameters. Prior constraints on simulation parameters consisted of the definition of the order of events (t3>t2>t1) and an initial effective population size between 10 and 100, where bottlenecks were simulated. Model scenarios were compared using a subset of 1% of all simulations, which matched the observed data best (Cornuet et al. 2014). Using a logistic regression approach, deviations from the summary statistics among the selected subset of simulations were used to predict the probability of a given scenario (Estoup et al. 2012). In order to reduce correlation among explanatory variables, summary statistics were transformed by linear discriminant analysis prior to logistic regression (Estoup et al., 2012).

Identification of genomic regions under selection

A commonly adopted approach to investigate genomic regions that may be under strong selective regimes is to look for $F_{\rm ST}$ outliers in population genomic datasets where several populations or locations have been sampled (e.g. Vandepitte et al. 2014; Lin et al. 2017; Leydet et al. 2018; Theodorou et al. 2018). This is based on the expectation that areas of reduced or elevated differentiation are subject to balancing and directional selection respectively (e.g. Lotterhos and Whitlock 2014). Other commonly used approaches to detecting selection in population genomic studies follow the expectation that directional selection decreases genetic variation in the genomic region of the selected site and balancing selection increases it (Oleksyk et al. 2010). Levels of genetic diversity can then be compared to the genomic background to infer candidate regions under selection (Hohenlohe et al. 2010; Bruneaux et al. 2013).

Approach 1: F_{st}-outlier approaches

To identify signatures of selection, we tested for elevated differentiation by first considering all pairwise comparisons between UK sampling sites and secondly for all pairwise UK-Le Havre comparisons. This allows distinction between patterns of differentiation across the UK and/or between UK and a continental population. Two approaches were used: a permutation approach (Bruneaux et al. 2013) and BayeScan (Foll and Gaggiotti 2008).

For the permutation approach, observed $F_{\rm ST}$ -values averaged across all pairwise comparisons of sampling sites were shuffled 1000 times and compared to the observed $F_{\rm ST}$ values averaged across SNP sites within 10 kb sliding windows. A 10 kb window size was chosen because linkage has been shown to decrease rapidly over these distances in *Bombus* (Sadd et al. 2015). To avoid single SNPs or RAD-tags driving the average across windows, a minimum SNP density of 3 SNPs was required across at least two independent RADtags for a window to be included in the analysis (Purfield et al. 2017; Jacobs et al. 2018). P-values were generated as the proportion of permutations being lower/higher than the observed estimates and corrected for multiple testing using a FDR approach implemented in the qvalue package in R (Storey et al. 2015). For the BayeScan approach (Foll and Gaggiotti 2008), the program was run using default parameters. BayseScan implements a basic regression model to differentiate between locus and population specific effects on the distribution of $F_{\rm ST}$ -values (Foll and Gaggiotti 2008). A likelihood ratio test is then used to assess if the population specific component is sufficient to explain the observed variation (no selection) or if a SNP specific component (selection) is supported (Foll and Gaggiotti 2008). This allows assessment of population specific demographic effects in contrast to the permutation approach, which identifies larger genomic areas with elevated $F_{\rm ST}$ values averaged across all sampling sites considered.

To reduce the rate of false positives (De Mita et al. 2013), genomic regions were only considered as outliers if they were identified using both the permutation and the BayeScan approach.

Approach 2: Distribution of polymorphic sites

Using the approach of Bruneaux et al. (2013), which evaluates the distribution of polymorphism across the genome against a random null-distribution and does not rely on allele frequency estimates, the proportion of polymorphic sites was calculated. This was performed across non-overlapping 10 kb sliding windows or across the length of the contig sequence in the case of genomic regions not placed within the assembled reference genome of *B. terrestris*. To generate a null-distribution 1,000 permutations were applied to the data and p-values were generated as the proportion of permutations being lower/higher than the observed estimates (Bruneaux et al. 2013). Only windows that had a minimum coverage of 100 bp were considered in the analysis (Cooper et al. 2004). For the analysis of low polymorphism, the hypergeometric test was used to derive the minimum coverage in base pairs needed across a window to obtain a probability below 5% of not sampling a SNP within a sliding window given our observed SNP density (Lentner, 1972; Fontanillas et al., 2010). In line with other studies, a false discovery rate (FDR) of 10% (q-value < 0.1) was applied (e.g. Krehenwinkel et al. 2015; Rane et al. 2015) as an appropriate balance between the false discovery rate and statistical power where large numbers of tests are involved (van den Oord 2008). Genes that were found within windows of significantly high/low polymorphism were considered for gene ontology analysis.

Gene ontology analysis

Gene ontology (GO) terms for the *B. terrestris* genome were obtained from the Ensembl database. Enrichment tests based on gene count (the 'classic' algorithm, Alexa et al. 2006) were conducted using the R package topGO (Alexa and Rahnenfuhrer 2016). These were assessed using Fisher's exact test and a minimum node size of 10 in order to prune our hierarchy from nodes with the support of less than 10 annotated genes, a frequently applied threshold (e.g. Ahrens et al. 2013; Rademacher et al. 2017). Correction for multiple testing (FDR < 5%, q-value < 0.05) was carried out using the qvalue package in R (Storey et al. 2015). Additionally, the 'weight' and 'elim' algorithms were used, which account for dependencies within the gene ontology hierarchies (Alexa et al. 2006). Here, multiple testing theory does not directly apply as tests are not independent and raw p-values ≤ 0.05 were considered as significant (Alexa and Rahnenfuhrer 2016). Genes within sliding windows that were identified as $F_{\rm ST}$ outliers or that showed significantly high or low polymorphism were subject to gene ontology analysis to investigate if outliers show significant enrichment for specific biological or molecular functions.

In order to classify SNPs as synonymous or non-synonymous, the annotation for the *B*. *terrestris* genome was obtained from the Ensembl database, which is assembled across 18 linkage groups.

Results

The range expansion of Bombus hypnorum into and across the UK

After the first record of *B. hypnorum* near Southampton in 2001 (Goulson and Williams 2001) more sightings followed across the south of the UK, with records increasing rapidly thereafter (Fig. 1). The sharpest increase in records was observed in 2009, where sightings increased more than four-fold compared to the previous year (likely to have been at least partly influenced by an increasing public awareness of the existence of *B. hypnorum* in the UK). The expansion was first in a north-eastward direction with some isolated records as far north as Hull in 2005 and Newcastle in 2007. The expansion then extended to the west with the first sightings of *B. hypnorum* in Cardiff in 2009, and in Plymouth in 2010.

RAD-seq data summary

Ion Torrent sequencing generated 15,513,192 raw reads for all sample pools. After mapping, quality filtering and the application of coverage thresholds 2,469,636 bp were covered in total among pools, representing $\sim 1\%$ of the expected genome size. This covers 40% of all expected cut-sites (57,157) of the enzyme used to digest samples (Xhol). Within each sampling site, the range of coverage was 152,225 to 953,719 bp (Table 1). After stringent filtering, 12,823 high confidence SNPs were identified in total, ranging from 464 to 3,695 within each sampling site. Mapping success was on average $6 \pm 3\%$ higher against the B. *impatiens* genome but the proportion of raw reads that were uniquely mapped per sampling site was highly correlated (rho 0.94, p < 0.001) between the *B. terrestris* and *B. impa*tiens genome, indicating minimal bias. We therefore used the *B. terrestris* genome to take advantage of annotated linkage groups, which has been shown to generally increase the power to detect selection across genomic regions compared with single site comparisons (Shafer et al. 2017). Across all genomic sites that were covered in all sampling sites (1,886, including non-polymorphic and polymorphic sites), the proportion of polymorphic markers varied from 0.006 (London) to 0.014 (Le Havre, Table 1), which fall within the range reported from other studies (e.g. Catchen et al. 2013). Deviation in the proportion of polymorphic markers between replicate pools was 0.001.

Allele frequency validation

In line with other studies we tested the consistency of our genotyping approach by comparing replicate allele frequency estimates (Anand et al. 2016; Guo et al. 2016; Dorant et al. 2019). A total of 3,065 SNP sites were covered in both Newcastle replicates and allele frequencies were highly correlated between them (rho 0.95, $p < 2.2 e^{-16}$) with an average difference in allele frequencies of 0.06. SNP allele frequencies within the individually barcoded Newcastle and Le Havre replicates compared to the pooled samples were high (rho 0.87, $p < 2.2 e^{-16}$ across 361 SNP sites for Newcastle 1, rho 0.97, $p < 2.2 e^{-16}$ across 507 sites for Newcastle 2, and rho 0.97, $p < 2.2 e^{-16}$ across 439 sites for Le Havre). The average allele frequency difference was 0.14 for the Newcastle 1 replicate, 0.07 for the Newcastle 2 replicate and 0.06 for the Le Havre pool in comparison to the respective individual data set.

Sampling site	Raw reads	Uniquely mapped reads	bp covered ^a	Number of SNPs (polymorph within populations)	rroportion of pory- morphic sites ^b	Allenc richness	Average heterozygosity ^c
Cardiff	1,832,315	1,306,780	655,120	2,139	0.012	1.22	0.15
Le Havre	2,350,459	1,726,971	323,767	2,251	0.014	1.26	0.20
Hull	2,394,796	1,508,923	593,141	2,384	0.008	1.16	0.12
London	1,159,972	619,972	245,067	851	0.006	1.11	0.12
Newcastle 1	2,788,910	1,781,942	953,719	3,695	0.012	1.23	0.18
Newcastle 2	2,466,615	2,012,501	322,409	1,929	0.011	1.21	0.20
Plymouth	714,223	397,903	152,225	464	0.01	1.18	0.13
Southampton	1,805,902	1,225,225	587,552	1,995	0.009	1.17	0.16
All UK populations	13,162,733	8,853,246	2,443,350	11,845	0.048	1.46	0.16
All populations	15,513,192	10,580,217	2,469,636	12,823	0.054	2	0.17

 Table 1
 Summary of RAD-seq data, including basic measures of genetic diversity

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^cAcross all SNPs found at the sampling site

Characterization of population structure

Average heterozygosity ranged from 0.12 in London and Hull to 0.2 in Le Havre and Newcastle (Table 1). Allelic richness ranged from 1.11 in London and 1.26 in Le Havre (Table 1). The deviation in average heterozygosity between the replicate pools was 0.02. Principal coordinate analysis on pairwise $F_{\rm ST}$ values was done for SNPs covered across all sample sites (101), a minimum of six sample sites (2166 SNPs) and a minimum of four sample sites (3475 SNPs) The pattern obtained was similar, with the Newcastle replicates clustered closely together with the Le Havre sampling site irrespective of the threshold of missing data that was applied (Fig. 3). $F_{\rm ST}$ -values were generally low (Table 2), and no significant isolation by distance was detected using a Mantel test (R^2 =-2.3⁻⁵, p=0.87). There was no significant difference in average $F_{\rm ST}$ between within-UK comparisons and UK–Le Havre comparisons (Mann–Whitney-Wilcoxon test, p-value>0.05). The average $F_{\rm ST}$ for within-UK comparisons was 0.023±0.008 and 0.016±0.007 for the UK–Le Havre comparison.



Fig. 3 Principal coordinate analysis on pairwise F_{ST} values between sampling sites for A) all SNPs covered in all sampling sites (101); B) all SNPs (2166) covered in at least six sampling sites (75%) and C) all SNPs (3475) covered in at least four sampling sites (50%); sample sites are abbreviated as follows: Le Havre (H), Southampton (S), London (L), Hull (Hu), Newcastle (N), Cardiff (C) and Plymouth (P)

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	Cardiff	Hull	Le Havre	London	Newcastle 1	Newcastle 2	Plymouth	Southampton
Cardiff	0.000	0.020,0.066	0.014, 0.039	0.012, 0.040	0.011, 0.040	0.014, 0.044	0.019, 0.056	0.010, 0.038
Hull	0.040	0.000	0.010, 0.029	0.009,0.066	0.012, 0.040	0.010, 0.030	0.010, 0.027	0.009, 0.043
Le Havre	0.026	0.019	0.000	0.006, 0.035	0.006, 0.015	0.003, 0.007	0.010, 0.034	0.007, 0.031
London	0.025	0.035	0.018	0.000	0.007, 0.025	0.006, 0.035	0.009, 0.057	0.006, 0.040
Newcastle 1	0.024	0.025	0.010	0.014	0.000	0.006, 0.015	0.012, 0.036	0.006, 0.025
Newcastle 2	0.028	0.018	0.005	0.018	0.010	0.000	0.009, 0.034	0.007, 0.030
Plymouth	0.036	0.018	0.020	0.032	0.023	0.019	0.000	0.009, 0.027
Southampton	0.023	0.023	0.017	0.018	0.014	0.017	0.017	0.000

Evaluation of colonization history

The continental reference sample (Le Havre), which we have assumed to be representative of the panmictic Western European population, was more likely to be the source of colonization than the simulated unsampled population in all cases (Fig. 4). Modelling results gave high support for multiple colonization events, indicating a direct colonization from continental Europe into Southampton, London and Newcastle and complex patterns of colonization within the UK (Fig. 4, Table 3). A bottleneck was supported for all populations with the exception of Southampton (Fig. 4, Table 3). The highest relative probability was observed for the colonization history of the London, Newcastle and Plymouth sites (Table 3). For the Cardiff, Hull and Southampton sites relative probabilities of the second most likely scenario were more similar and confidence intervals were overlapping (Table 3). In the cases of Cardiff and Southampton, the two top scenarios differed in their support for a dual colonization history, whilst for the Hull site the probability of a colonization from London was closely followed by the probability for a direct colonization from the Western continental population (Table 3).

Detection of selection

Of all sites covered (2,469,636 bp), 136,626 bp fell within coding sequences, of which 557 were polymorphic. Of these polymorphisms, 346 were non-synonymous and 211 synonymous. A total of 75 F_{ST} -outliers (66 under directional selection and 9 under balancing selection) were identified by BayeScan. The permutation approach identified 8 windows with significantly elevated F_{ST} -values (6 for within UK comparisons and 2 for UK-Le Havre comparisons) and 11 windows with significantly low F_{ST} -values (3 for within UK comparisons and 8 for UK-Le Havre comparisons) (Fig. 5). The analysis of patterns of polymorphism across the genome revealed 1,219 windows with significantly elevated polymorphism (Fig. 6). Given an observed SNP density of 4.5 per kB, the minimum coverage in base pairs required to obtain a probability below 5% of not detecting any SNP within a sliding window was 645 bp, as shown by the hypergeometric test. This resulted in a total of 383 windows exhibiting significantly reduced polymorphism (Fig. 6).

Genes that showed evidence of directional selection from both approaches and all three methods (i.e. $F_{\rm ST}$ -outlier permutation, BayeScan *and* the polymorphism approach) were the protein vestigial (Le Havre-UK comparison), the circadian locomotor output cycles protein (kaput), and one gene important in signal transduction (serine/threonine-protein kinase NLK), both from within-UK comparisons (Table 4). None of the outliers identified by BayeScan to be under balancing selection fell within windows of low $F_{\rm ST}$ in any of the data sets.

GO analysis

GO analysis revealed areas of high polymorphism that were significantly associated with biological processes, including regulation of transcription and gene expression, signaling and developmental processes (Table 5a). Further, areas of high polymorphism were significantly associated with the molecular functions of protein and sequence specific DNA binding, DNA binding transcription factor activity and zinc ion binding Fig.4 A) Geographic representation of the DIYABC consensus model B) Schematic consensus model, ► derived by combining the scenarios with highest likelihood for each UK sampling site. Three independent colonization events from the continental reference site, Le Havre (H), one to Southampton (S) one to London and one to Newcastle (N) are apparent. Further, founders from close UK sampling sites (London (L), Hull (Hu), Southampton (S) and Newcastle (N)) are also involved in the establishment of populations. Note that, whilst the succession of events implies a certain timeline, time is not explicitly evaluated here and not represented by the length of connecting lines. *representing the continental European reference population

(Table 5a). Areas of significantly low polymorphism were associated with the molecular functions of transmembrane signaling receptor activity and extracellular ligand-gated ion channel activity (Table 5b). GO analysis did not reveal any significant associations for molecular functions or biological processes for windows identified as outliers within any of the $F_{\rm ST}$ data sets.

Discussion

The colonization of the UK by *Bombus hypnorum* is an important and interesting counterexample to the reported widespread decline in bumblebees across the northern hemisphere (Goulson et al. 2008). Here we report (i) evidence for multiple entries into the UK (ii) similar levels of genetic diversity in the sampled UK and continental population, despite indications of initial bottlenecks in some sites, and (iii) preliminary evidence of selection in some genomic regions.

The population expansion of B. hypnorum into and across the UK

For the first time, we report population genomic evidence that supports multiple colonization of the UK by *B. hypnorum*. Collated records from the BWARS database indicate an initial colonization of the UK in the south, with a subsequent rapid spread north and east and a later spread westwards. Our modelled scenarios generally support the colonization pattern suggested by the BWARS database although some sample sites were founded from multiple sites. There is a high likelihood of migration into multiple sites from continental Europe across the south and north-east, combined with ongoing and rapid migration from neighboring sites already colonized. In bumblebees, queens are the founders of new nests and therefore drive the effective dispersal of the species (Lepais et al. 2010). Our results suggest that both jump-dispersal over a longer distance outside the native range, as well as diffusion dispersal, the gradual dispersion over shorter distances (Pielou 1979), are playing a role in explaining the colonization pattern of B. hypnorum queens into the UK. A pattern of multiple founders from both local and continental sources is supported further by patterns of genetic differentiation. For example, the Newcastle sample clustered most closely with Le Havre in a PCoA on pairwise F_{ST} (Fig. 3), which again suggests an independent introduction from the Western continental population rather than colonization from within the UK only.

Evidence for loss of genetic diversity in colonizing populations

Whether or not colonization is accompanied by a loss of genetic diversity is important in terms of understanding both the colonization event itself, and the potential





Legend

C Cardiff	L London	S Southampton	\rightarrow direction of colonization
$\overset{\smile}{\mathbb{H}}$ Le Havre	Newcastle	(X) unknown continental	→ bottleneck
Hu Hull	P Plymouth	C common ancestor	

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Table 3 relative prot	vabilities and confidence intervals for the top	two scenarios in the DIYABC ana	alysis for each test site	
Test site	Most likely source of colonization	Relative probability [CI]	Second most likely scenario	Relative probability [CI]
London	Le Havre bottlenecked	0.92 [0.75,1.00]	Hull bottlenecked	0.0660 [0.00,0.24]
Cardiff	Combined colonization Hull-Newcastle bottlenecked	0.5811 [0.14,1.00]	Newcastle bottlenecked	0.4189 $[0.00, 1.00]$
Hull	London bottlenecked	0.5463 [0.35, 0.74]	Le Havre	0.4537 [0.26,0.65]
Southampton	Le Havre	0.6065 [0.13,1.00]	Combined colonization Le Havre-Hull	0.3917 [$0.00, 0.86$]
Newcastle	Combined colonization Le Havre-South- ampton bottlenecked	1.0000 [1.00,1.00]	NA	0.0000 [0.00,0.00]
Plymouth	London bottlenecked	$0.9661 \ [0.92, 1.00]$	Le Havre	0.0339 $[0.00, 0.08]$



Fig. 5 Distribution of average F_{ST} -values across linkage groups across all pairwise comparisons within the UK (points) and between the UK and Le Havre (crosses) respectively. Genomic regions showing significantly high F_{ST} -values based on permutation tests are highlighted in red and regions with significantly low F_{ST} -values are highlighted in blue



Fig.6 Distribution of polymorphism across linkage groups. Genomic regions showing significantly elevated polymorphism based on permutation tests are highlighted in red and areas of significantly reduced polymorphism in blue. Note the different y-axis scale in the NW linkage group

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GeneID	Description	Chromosome	Exon count	Comparison	Function
100,651,612	Protein vestigial	LG B12	9	Le Havre – UK	Wing formation/ wing specific gene expression
100,650,446	Circadian locomotor output cycles protein kaput	LG B09	12	UK populations	Circadian clock
100,651,912	Serine/threonine-protein kinase NLK (nemo-like kinase)	LG B03	11	UK populations	Involved in signal transduction pathways that regu- late growth factor response, proliferation, apoptosis and innate immune response
110,119,491	Uncharacterized LOC110119491	LG B08	2	UK populations	Unknown

G0.ID Term Annotated Significant Expected Class (A) (A) $(0.0)50,794$ regulation of cellular process 373 118 86 0.000 BP GO:0,050,794 regulation of cellular process 373 119 87.85 0.000 GO:0,050,794 regulation of cellular process 371 19 87.85 0.000 GO:0,050,759 regulation of cellular process 371 19 87.85 0.000 GO:0,005,355 regulation of gene expression 37 19 8.53 0.000 GO:0,005,355 regulation of transcription, DNA-templated 144 51 33.2 0.000 GO:0,007,154 cell communication 230 73 53.03 0.000 GO:0,007,255 multicellular organism development 230 73 53.03 0.000 GO:0,007,255 multicellular organism development 211 12 24.71 0.003 GO:0,008,770 DNA-binding transcription factor activity 24 24.71 <td< th=""><th>Table 5 across ; elevate(</th><th>Gene ontology enri all sampling sites. Re 1 (A) and reduced (B</th><th>chment analysis of Molecular Function (MF) and E sults are shown where at least two of the three algo) polymorphism.</th><th>siological Process rithms used ('clas</th><th>ses (BP) for all re isic', 'elim' and '</th><th>sgions showing (weight') support</th><th>significantly ele ted significance</th><th>vated/reduced p for regions with</th><th>olymorphism significantly</th></td<>	Table 5 across ; elevate(Gene ontology enri all sampling sites. Re 1 (A) and reduced (B	chment analysis of Molecular Function (MF) and E sults are shown where at least two of the three algo) polymorphism.	siological Process rithms used ('clas	ses (BP) for all re isic', 'elim' and '	sgions showing (weight') support	significantly ele ted significance	vated/reduced p for regions with	olymorphism significantly
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$ \begin{array}{cccccccccccccccccccccccccccccccccccc$		GO:0,010,468	regulation of gene expression	151	53	34.82	0.000200	0.458540	0.000200
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GO:0,005,230 extracellular ligand-gated ion channel activity 16 8 3.58 0.013		GO:0,004,888	transmembrane signaling receptor activity	37	17	8.28	0.000940	0.000940	0.000940
		GO:0,005,230	extracellular ligand-gated ion channel activity	16	8	3.58	0.013090	0.013090	0.013090

Evolutionary Ecology

Significant p-values are highlighted in bold

evolutionary constraints and vulnerability of the newly established populations (Dlugosch and Parker 2008), especially in haplodiploid species where effective population size is reduced (Lester and Selander 1979). Here, we observed generally comparable proportions of polymorphic markers and measurements of heterozygosity within the continental Le Havre and UK locations (Table 1). This is consistent with a previous report that found no evidence for the loss of genetic diversity in a single UK population (Crowther 2017). The levels of genetic diversity observed here in *B. hypnorum* are similar to other long-established bumblebee populations in North America based on SNP data (Lozier 2014). This is despite the fact that DIYABC analysis suggests that bottleneck events have occurred during the establishment of *B.hypnorum* in the UK (Fig. 4). However, the likely occurrence of multiple introductions through the presumed successive influx of incoming dispersing queens, the dynamic patterns of migration within colonized UK sites and the rapid expansion will mitigate a loss of genetic diversity (Nei et al. 1975; Pannell and Charlesworth 2000; Zenger et al. 2003; Dlugosch and Parker 2008).

As we sampled only one continental location, the level of variation in genetic diversity across potential source populations is unknown and so could be under-represented in our study. Studies on other bumblebee species that are widespread and abundant across continental Europe (*B. terrestris* and *B. pascuarum*) did not reveal population structuring on a continental scale (Estoup et al 1996; Widmer and Schmid-Hempel 1999). This suggest that across continental Europe there are no significant barriers to gene flow for bumblebee species with ample dispersal abilities, consistent with other studies suggesting that connectivity is generally high for mainland populations of bumblebees (Lozier 2014). It is therefore likely that population structuring is minimal across the coast of north-western continental Europe in *B. hypnorum*. Indeed, a very recent study did not find population structure across populations of *B. hypnorum* in Belgium (Maebe et al. 2019).

Jones and Brown (2014) used an indirect approach of diploid male production as a preliminary assessment of genetic diversity of *B. hypnorum* in the UK. They reported lower than expected genetic diversity within the UK compared to continental European populations, in contrast to our results here. However, that study focused exclusively on samples from the London area, and here we also report this area as having the lowest diversity of all of the locations sampled (both in terms of heterozygosity and in polymorphic sites), raising the possibility that the London site may be less representative of the wider UK population.

Our finding of no major loss of genetic diversity in a successfully colonizing species is consistent with the majority of cases documented (Roman and Darling 2007; Dlugosch and Parker 2008). However, there are other contrasting examples, such as the colonization of North America by the solitary bee *Lasioglossum leucozonium*, which was likely initialized by the introduction of a singly-mated female (Zayed et al. 2007). Different patterns of loss of genetic diversity through founding events are likely explicable by the severity and length of periods of reduced population size, consistent with established population genetic theory. This suggests that despite initial small population sizes in the UK as indicated by the detection of bottlenecks in the DIYABC analysis, founding populations expanded quickly, which in combination with the likely ongoing independent colonization events, minimized the loss of genetic diversity. This is further supported by a very recent study using microsatellites that reports no significant increases in diploid males in the UK population (Brock et al. 2021).

Evidence for selection

The limitation in the identification of signatures of selection across the sampled sites using a RAD-seq approach is that only a portion of the genome is screened (here 1%). so many loci under selection are likely to be missed (Tiffin and Ross-Ibarra 2014; Lowry et al. 2017). Further, we used the genome of a different species to infer functionality of genomic regions. Although this is common practice when working with non-model species (Shafer et al. 2017), the annotation is likely to be incomplete and areas of high divergence to the reference genome may fail to be annotated. The complex demographic history and rapid expansion may also leave genomic signatures similar to those expected under selection (Excoffier and Ray 2008; Li et al. 2012). By taking a stringent approach and only highlighting those genes identified from all our approaches (two F_{ST} -outlier approaches and genomic regions showing significantly elevated or low polymorphism) three genes were identified as putatively showing signatures of selection. Only one gene, encoding for the protein 'vestigial' and involved in wing formation, showed significantly elevated differentiation between Le Havre and UK sites. Geographic variation in wing shape and size is common in insects (Hoffmann and Shirriffs 2002; Kandemir et al. 2009) and the establishment of geographic clines has been documented within short time scales (Gilchrist et al. 2001), which may also be the case here. Differences in wing morphology may also relate to flight and dispersal abilities. Spatial sorting theory predicts that dispersal ability is a trait under strong selection during range expansions (Shine et al. 2011; Berthouly-Salazar 2012). This is explained by the expectation that those individuals with the highest dispersal abilities will be spatially assorted at the expanding range front. This creates a positive feedback loop as expansion continues and accelerates the speed at which new areas are colonized (Berthouly-Salazar 2012).

Among UK-UK site comparisons, two further genes were highlighted as being potentially under strong directional selection: the CLOCK gene, which has an important role in the regulation of circadian rhythms (Darlington et al. 1998) and has been linked with adaptive responses to environmental conditions across a range of taxa, including invertebrates (Tauber and Kyriacou 2005; O'Malley and Banks 2008; Liedvogel et al. 2009) and serine/threonine protein kinase NLK, which is associated with innate immune function and apoptosis (cell death) (Mirkovic et al. 2002; Li et al. 2014). A serine/threonine protein kinase region was also identified as an $F_{\rm ST}$ – outlier by Theodorou et al. (2018) in a recent study using a RAD-seq approach in *Bombus lapidarius*, where it was suggested as a signature of an adaptive response to increasing urbanization. Colonizing *B. hypnorum* suffer higher prevalence of highly virulent parasites than native species (Jones and Brown 2014; Lloyd et al. unpublished data), which might also be explanatory.

Further, GO analysis identified areas of significantly high polymorphism were associated with biological processes or molecular functions, largely regulating transcription and gene expression (Table 5). This highlights research into the role of genetic variation in regulatory genes in the adaptability of *B. hypnorum* to new locations as a potentially important area for further studies. The GO analysis of genes within genomic regions of low polymorphism revealed a significant association with ion channel and signalling receptor activity (Table 5) and the investigation of the adaptive role of these genomic regions similarly warrants further research across a wider group of *Bombus* species.

Why now?

Our study has revealed a probable pattern of multiple colonization events from western continental Europe of *B. hypnorum*, in addition to ongoing gene flow and spread from within established UK sites. We also report evidence of some genomic areas that may show signals of directional selection. Given the likely pattern of colonization, a key question is why this species has been such a successful and successive colonizer now, when it has been present on continental Europe for many decades prior to the turn of the twenty-first century. Answers to this can only remain speculative but our study gives some potential insight by suggesting a pattern of multiple, possibly ongoing, colonization routes with no evidence of a reduction in genetic diversity. Following the colonization success framework of Facon et al. (2006) our results highlight a possible role of 'migration change' where barriers to dispersal have recently and relatively suddenly opened. Possible changes to anthropogenic transport routes are one potential explanation. Anthropogenic rates of trade and transport are subject to continuous growth (Hulme 2009) and although there have been no obvious significant changes over the period in question, increased opportunities for assisted introduction may have facilitated the colonization of the UK. Shifts in climatic conditions are another potential explanation, which may facilitate dispersing queens reaching the UK. While temperature changes alone are unlikely to be responsible, given that the UK sits well within the current climatic range of *B. hypnorum*, shifts in wind patterns (e.g. Hu et al. 2016; Weber et al 2018) are among these plausible scenarios. There is less compelling current evidence for a role of 'environmental change'. While there are noted ecological differences between *B. hypnorum* and other UK species, both in nesting site preferences and also both habitat and foraging associations (Crowther et al. 2014), there is little evidence to date of any marked change in availability of either nesting sites or habitat over the period in question. Neither is there any evidence for 'enemy release' occurring, from recent comparative study of *Bombus* parasite communities (Jones and Brown 2014). Our preliminary results highlighting genes already suggested to be involved in adaptive responses to urbanized environments in similar or other insect groups lend some traction to the third 'evolutionary change' scenario as a contributing factor underpinning the remarkable expansion of this species. Such adaptations to anthropogenically modified habitats within the native range, have been put forward as an important feature to promote range expansions (Hufbauer et al. 2012). Finally, biogeographic factors could play a role: with declines of native species on an island (some UK bees have been completely extirpated and many others are declining), it might be expected that novel species might colonize. Further investigation into the potential drivers of this population expansion, particularly focusing on migration and/or evolutionary changes, are likely to yield key insights to our understanding of these recent population changes. Notably, establishing whether these are ubiquitous signals across a much wider taxonomic range, or whether this particular system has been successful because it is an exception to the norm will give important insights into the rapid changes in distribution and abundance of species currently being witnessed.

Acknowledgements We are grateful to the University of Plymouth for funding this project and to our colleagues for insightful comments on earlier drafts of this manuscript.

Author contributions MEK, JSE and MJFB conceived the project and secured funding; KL carried out the sampling; KL, CMB and MK generated the raw data; JVH conducted the analysis in discussion with MEK and JSE; JVH, MEK and JSE drafted the manuscript, with input from the other authors; MEK led the project.

Funding Funding for this project was provided by the University of Plymouth.

Data accessibility Raw reads generated in this study are available on GenBank (SUB9082181).

Declarations

Conflict of interest The authors declare no conflicts of interest.

Consent to participate Not applicable.

Consent for publication Not applicable.

Ethics approval Not applicable.

Code availability A custom python script used for variant calling is available on GitHub: https://github.com/ janavanhu/genotyping_pools_from_sync

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