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Behavioural modes in butterflies: their implications for movement and searching behaviour

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Animals move in 'modes' where movement patterns relate to specific behaviours. Despite much work on the movement of butterflies, their behavioural modes are relatively unexplored. Here we analysed the behaviour of the model butterfly species the meadow brown, *Maniola jurtina*. We identified modes in both sexes and across habitats varying in resource density. We found that, in nectar-rich habitats, males had more diverse behaviour than females, engaging in a unique 'high-flight' mode associated with mate search, whereas females were primarily nectaring or inactive. In nectar-poor habitats, both sexes were similar, switching between flight and inactivity. We also identified the movement parameters of the modes, finding that, for both sexes, movements associated with nectaring were slower and more tortuous and, for males, the mode associated with mate searching was straighter and faster. Using an individual-based random-walk model, we investigated the effects of

behaviour on movement predictions by comparing a mode-switching model with a version including intraspecific variation and another assuming homogeneity between individuals. For both sexes, including modes affected the mean and shape of the displacement rate compared to models assuming homogeneity, although for females modes increased displacement 1.5 times while for males they decreased it by a third. Both models also differed substantially from models assuming intraspecific variation. Finally, using a new model of search behaviour we investigated the general conditions under which individuals should engage in an exclusive search for host plants or receptive females. Parameterized for *M. jurtina*, the model predicted males should engage exclusively in mate search, but females only when searching is very efficient. The model provides a framework for analysing the searching behaviour of other butterfly species.

Keywords: *Maniola jurtina*, meadow brown, motivation, movement

A fundamental aspect of the behaviour of an animal is the way it moves through its environment. Movement is evaluated from several standpoints (sensu Tinbergen 1963) varying from the mechanistic or biomechanical (e.g. animal locomotion; Alexander, 2003) to considerations of adaptive purpose (e.g. optimality; Charnov, 1976). Current research in movement ecology intersects these areas (Holyoak, Casagrandi, Nathan, Revilla, & Spiegel, 2008), linking the observed movement of individuals to the motivational states that underpin them (Nathan et al., 2008). A key assumption of much recent modelling is that animals switch between distinct movement patterns, often referred to as ‘modes’, as a result of the local environment and their motivation (Fryxell et al., 2008; Morales & Ellner, 2002; Morales, Haydon, Frair, Holsinger, & Fryxell, 2004; Skalski & Gilliam, 2003). Here modes refer to temporally and spatially correlated movement patterns adapted to achieving a specific

goal (e.g. foraging). Identifying modes offers many benefits: first, it allows quantification of the features of a behaviour (Jonsen, Myers, & James, 2006; Weimerskirch et al., 2002); second, it links behaviours to the distribution of individuals across landscapes (Börger, Dalziel, & Fryxell, 2008; Singh, Börger, Dettki, Bunnefeld, & Ericsson, 2012); and third, the optimality of the movement patterns for their inferred purpose can be examined (Avgar, Kuefler, & Fryxell, 2011; Dias, Granadeiro, & Palmeirim, 2009; Louzao, Wiegand, Bartumeus, & Weimerskirch, 2014). However, linking movement modes with their associated behaviours is challenging, as the accompanying behaviour is not always observed, and internal motivations are hidden.

There has been a dramatic increase in the collection of movement data (Williams et al., 2020), owing to remote technologies such as global positioning systems (GPS; Hebblewhite & Haydon, 2010; Seidel, Dougherty, Carlson, & Getz, 2018). A challenge with these data is that behaviours accompanying movements are not typically recorded. Behavioural modes, therefore, must be inferred through statistical techniques (Patterson et al., 2017; Schick et al., 2008), such as change point analysis (Killick and Eckley 2014) or state space modelling (Patterson et al. 2008), that detect behavioural states in a time series of coordinates (Gurarie et al. 2016). However, the method for tracking butterflies, a model group for the study of movement and dispersal (Stevens, Trochet, Van Dyck, Clobert, & Baguette, 2012; Stevens, Turlure, & Baguette, 2010), is unusual, as movements have often been recorded by directly observing individuals over short timescales (Odendaal et al., 1989; Root & Kareiva, 1984; Schultz, 1998; Schultz, Franco, & Crone, 2012; Turchin, 1991). An advantage of this approach is that behaviours are recorded concurrently with movement data, and can be categorized simply (Dover, 1989), generating contemporaneous movement and behavioural information. This allows evaluation of the effect of observed behaviours on movement rates, rather than inferring behaviour from movement data.

78 Previous studies have typically investigated movement modes in taxa larger and longer lived
79 than insects. In the Artiodactyla, movements transition broadly between encamped and
80 exploratory modes (Fryxell et al., 2008; Morales et al., 2004), further refined to bedding and
81 foraging (Franke, Caelli, & Hudson, 2004), predator avoidance (Forester et al., 2007) and
82 seasonal migration (Singh et al., 2012). Similarly, for pinnipeds, movement modes have been
83 identified for foraging at different depths (McClintock, London, Cameron, & Boveng, 2017),
84 in different seasons (Breed, Jonson, Myers, Bowen, & Leonard, 2009), and for engaging in
85 directed and undirected travel (Gurarie, Andrews, & Laidre, 2009). For butterflies, changes in
86 modes have often been associated with transitions between patches of varying habitat quality,
87 where movement rates change in response to resource densities (Fownes & Roland, 2002;
88 Odendaal et al., 1989; Ovaskainen, Luoto, et al., 2008; Schtickzelle, Joiris, Dyck, &
89 Baguette, 2007). Less explored is how butterfly movements vary within contiguous habitats
90 in response to motivation, and the consequences this has for movement rates and the fitness
91 of individuals.

92 It has long been recognized that butterflies engage in behavioural modes targeted at specific
93 purposes (Shreeve, 1992). Dennis and Hardy (2007) observed pierid species performing
94 foraging or directed flight patterns in response to habitat quality, and studies using harmonic
95 radar show butterflies engaging in distinct foraging or dispersive flights (Cant, Smith,
96 Reynolds, & Osborne, 2005). There is also much research on the sex-specific behaviours of
97 butterflies (Scott, 1974; Wiklund, 2003). For example, Brakefield (1982a) noted meadow
98 brown, *Maniola jurtina*, males engaging in patrolling behaviours, seeking out females on
99 sustained flights. Similarly in other satyrids, males are known to switch between territorial
100 and patrolling behaviours (Shreeve, 1984; Takeuchi, 2010; Wickman, 1985; Wiklund, 2003).
101 Thus, butterflies appear to perform distinct modes related to specific goals and these may be
102 consequential for understanding the movement of individuals and their distribution in an

environment. Including behavioural variation in models of butterfly movement is known to affect predictions of movement rates (Evans et al., 2020b, 2020a); however, the way behavioural differences are implemented may influence predictions. In many random-walk or diffusion approaches individuals, at some level, are considered identical (Gurarie, Anderson, & Zabel, 2009); thus, within a given habitat, movement observations may be pooled (Evans et al., 2020b; Schultz & Crone, 2001). However, other approaches maintain behavioural variation between individuals within the same habitat (Brown & Crone, 2016; Korösi, Örvössi, Batáry, Kövér, & Peregovits, 2008). Not well considered thus far is the effect of implementing behaviour through state switches, as has been applied in many other taxa (Morales et al., 2004; Patterson et al., 2017).

Here we aimed to explore butterfly movement behaviour through identifying and describing the behavioural modes of the model butterfly *M. jurtina*. To achieve this we utilized a large data set of both movement and behaviour, collected within areas of varying habitat quality (Evans, Sims, et al., 2019). Specifically, we aimed to (1) identify behavioural modes in the sexes and between habitats of varying resource density, (2) quantify the movement parameters associated with the modes, (3) demonstrate the consequences of behavioural modes for the distribution of individuals across a landscape using an individual-based random-walk model and (4) develop a model, from first principles, to explore the optimality of exclusive search modes for reproductive resources. We focused especially on the effect of behavioural modes in high-quality habitat as this is far less explored than the effects of habitat quality on movement.

<H1>Methods

<H2>Species

Maniola jurtina is a common butterfly found throughout the British Isles and much of Europe (Fox et al., 2015). It is a characteristic grassland specialist (Van Swaay et al., 2019), with larvae feeding on common grasses and herbs (Ouin, Martin, & Burel, 2008) and the adults obtaining nectar from a variety of flowers (Brakefield, 1982a; Dennis, 1992; Lebeau, Wesselingh, & Van Dyck, 2017). The species is univoltine and typically on the wing between June and September (Thomas 2010). The males emerge first (Brakefield, 1982b; Scali, 1971) and are more active flyers than the females, spending extra time in flight searching for receptive females (Brakefield, 1982a; Evans et al., 2020a; Evans, Sibly, et al., 2019a). The females are monandrous and typically mate quickly after emergence (Dowdeswell, 1981). When choosing host plants they are relatively unselective (Delattre et al., 2010), although they show preferences for short grasses (Lebeau, Wesselingh, & Van Dyck, 2015). The adult life span in the British Isles is 5–12 days although can be as long as 20 days (Brakefield, 1982b), with survival duration probably reflecting the amount and quality of nectar resources (Evans, Sibly, et al., 2019b; Lebeau, Wesselingh, & Van Dyck, 2016a).

<H2>Movement and behavioural data

An open-access data set of butterfly movement was analysed (Evans, Sims, et al., 2019) and as methods for this data collection are provided elsewhere we here provide only a brief description. Butterflies were followed opportunistically with movement and behaviour recorded simultaneously. Following a standard approach (Odendaal et al., 1989; Schultz, 1998; Turchin, 1991), movements were recorded by laying marker flags every time the butterfly alighted or every 15 s during continuous flight. Observations ceased after 10 min or after either 15 or 20 flags had been laid (15 flags in 2018 and 20 in 2017). The coordinates of

the flags were then retroactively mapped using a high-grade global navigation satellite system receiver (Arrow 200 RTK GNSS, Eos Positioning Systems, Inc., Terrebonne, QC, Canada). During the observations, behaviours were recorded continuously by categorizing behaviour into flying, nectaring (taking nectar from flowers), basking (open wings and stationary), inactive (closed wing and stationary) or ovipositing (Dover 1989). Timing of behaviour was recorded accurately using a bespoke android phone app developed for the project.

Observations were relatively balanced between the sexes (184 ♀, 242 ♂), with most observations taking place in nectar-rich habitats (rich: 321; poor: 105). Data on individual flight tracks were collected over 72 days during the summers of 2016 (July–August), 2017 (June–September) and 2018 (June–July), at four sites in the south of England: North Farm in Oxfordshire (51°37'N, 1°09'W), Jealott's Hill Farm, Berkshire (51°27'N, 0°44'W), the University of Reading (51.4414° N, 0.9418° W) and Sonning Farm, Berkshire (51°28'N, 0°53'W). Three of the sites were agricultural farms that had implemented agri-environment schemes and consisted of a mixture of arable fields, open meadows and nectar-rich field margins, while the fourth consisted of areas of meadow within the grounds of the University of Reading campus. Data were labelled dichotomously as either nectar rich or nectar poor, with nectar-rich areas consisting of grasslands with a variety of wildflowers while the nectar-poor sites were mowed grass with very few flowers. Hourly air temperature was collected from local meteorological stations and mean solar radiation during observations (recorded every 10 s) from dataloggers (HOBO pendant, Tempcon Instrumentation, Arundel, U.K.).

<H2>Ethical note

Permissions were obtained from landowners for all sites visited during observations (The Earth Trust, Syngenta Jealotts Hill, the University of Reading, Sonning Farm University of

Reading). All observations took place in the field and no butterflies were handled. The methods applied for observing butterfly movement have been demonstrated to have no observable impact on behaviour (Root & Kareiva, 1984)

<H2>*Statistical analysis*

To identify behavioural modes, the analysis was conducted in two stages. First, data were collated into time budgets and a clustering approach was applied to group butterflies performing similar behaviours across an entire observation. This we consider as representative of a behavioural ‘mode’. In this first stage, observations from both sexes and all habitat types were pooled and sex and habitat were used as predictors of cluster group identity in a multinomial regression. This first stage identified that, as expected, sex and habitat strongly predicted cluster grouping (see Results). Consequently, in the second stage, observations were split by sex and habitat type and a separate cluster analysis was performed to evaluate groupings in each sex*habitat combination. Identifying clusters in the nectar-poor habitat allowed us to compare responses to habitat quality with those found in the literature, although our analysis primarily focused on the modes of butterflies within nectar-rich habitats.

Silhouettes (Rousseeuw, 1987) were used to identify the number of clusters, ranking the proposed number by comparing the distances of objects contained within a cluster to the distance of the nearest neighbour of an adjacent cluster. Implementations are available in the R package ‘cluster’ (Maechler, Rousseeuw, Struyf, Hubert, & Hornik, 2019). This provided an objective method to select the number of behavioural clusters in the data. With the number of cluster groupings selected, *K*-means clustering (Hartigan & Wong, 1979) was applied. For comparisons in stage 1, air temperature was introduced as a covariate, as insolation data were

unavailable in nectar-poor regions. For stage 2, however, within the nectar-rich habitat, insolation was found to be a far superior predictor of clustering than air temperature (lower Akaike information criterion) and was used instead.

To compare step distances and turning angles between clusters within nectar-rich habitats, Tukey's test for comparing individual means was used for the step distances (Tukey, 1949). Wallraff rank sum tests of angular distance were used to compare differences in turning angles. Step distances were log transformed to meet the assumptions of Tukey's test.

Multinomial regression was carried out using the package 'nnet' (Ripley, Venables, & Ripley, 2016), Silhouettes were produced using 'factoextra' (Kassambara & Mundt, 2017), and Wallraff rank sum tests using 'circular' (Agostinelli & Lund, 2017) all within R 3.6.1 (R Core Team, 2019).

<H2>Random-walk models

To explore the effect of behavioural modes on movement rates and the distribution of individuals, a simple individual-based random-walk model was developed. The model predicted the daily displacement from a fixed starting point after 8 h (28 800 s) of simulated time. Three versions were produced; all were sex specific but differed in their treatment of behaviour. The 'pooled' model simulated behaviour without reference to any behavioural variation; this we considered typical of random-walk or diffusion approaches where individuals are considered identical (Gurarie, Anderson, et al., 2009) and observations within a habitat type often pooled (Evans et al., 2020b; Schultz & Crone, 2001). The 'mode' version was a state switch model (Morales et al., 2004; Patterson et al., 2017) that included transitions between modes and variation in step distances, turning angles and the proportion of time in flight for the behavioural modes. Finally, in the 'intraspecific' model, variation

between individuals in movement propensity is seen as a fixed trait (Korösi et al., 2008) and the amount of time in flight matched the proportions observed. The models contain variation from different sources, the ‘pooled’ model from stochasticity in steps and turns, the ‘mode’ model additional stochasticity in choice of behavioural mode and the ‘intraspecific’ model fixed individual variation in flight propensity.

In the pooled model, butterflies moved by selecting steps and turning angles from all data in nectar-rich habitats. Movement occurred for a ‘flight time’ which was the mean of the observed proportion of time in flight multiplied by the total simulated time. For example, if butterflies spent on average 20% of their time in flight then total simulated flight time would be $0.2 \times 28\,800\text{ s} = 5760\text{ s}$. For every step, the mean duration of the step distances was subtracted, and movement stopped when all butterflies had run out of flight time. For the mode model, butterflies selected behavioural modes in proportion to those observed in the data. Each mode had a cluster-specific step distance distribution, turning angle distribution and flight time (Tables S1 and S2). To match the timescale of the data collection, modes switched every 10 min of simulated time, with the frequency of each mode proportional to that observed in the data. The flight time in each of the 10 min was the average proportion of time in flight for that cluster multiplied by 600 s and this was repeated until the total simulated time had elapsed. In the intraspecific model, butterflies drew the proportion of time in flight from observations but moved using pooled steps and turning angles. The pooling at this stage was used because not all butterflies had sufficient steps to generate appropriate individual turning angle and step distributions.

To compare the models, for each sex and model type combination, 5000 butterflies were initialized at the centre of a 2 x 2 km landscape and the model was run for the simulated day. The landscape was made sufficiently large to avoid any edge effects. At the end of the run Euclidean distance from the start location for each butterfly was then recorded, thus

representing the total displacement and the change in the distribution of the butterflies in the habitat.

The model was built in NetLogoR (Bauduin, McIntire, & Chubaty, 2019), a recently developed set of individual-based functions inspired by the NetLogo language which can be used for developing individual-based models within R (code is available at DOI: 10.17632/mm2skm8f6j.1). Turning angles were simulated using the ‘circular’ package (Agostinelli & Lund, 2017).

Exclusive search model

A model was derived from first principles to explore the utility of searching behavioural modes for both sexes. The model conceptualized a trade-off between exclusively searching for the resources associated with reproductive fitness (e.g. host plants, receptive females) and finding these resources as a by-product of normal, lower net-energy expenditure, behaviours of foraging and inactivity. Specifically, the model aimed to explore how much time individuals should dedicate to an exclusive search mode given (1) the change in resource over time, (2) the energetic cost of search behaviour and (3) the relative effectiveness of exclusive search over normal behaviours. It is assumed that when in exclusive search mode butterflies trade life span for resources by consuming no nectar and thus use reserves, resulting in reduced life span (Evans, Sibly, et al., 2019b; Lebeau et al., 2016a; Vande Velde & Van Dyck, 2013). It is also assumed that butterflies can maintain net energy balance in inactive/foraging modes by replenishing expended energy with nectar sugar and becoming inactive to reduce metabolic rate (Lebeau, Wesselingh, & Van Dyck, 2016b; Niitepold, 2010; Niven & Scharlemann, 2005). Thus, lifetime energy use can be represented as:

$$L_e = E_s T_s$$

where L_e = lifetime net energy use, E_s = energetic cost of search (J/s), T_s = time in search (s)

This can be converted to a reduction in survival time by multiplying by a factor, A , that relates energy loss to survival time. Combining E_s and A creates variable A' and the equation for predicting life span is:

$$LS = LS_n - A'T_s$$

where LS = life span (s) and LS_n = maximal life span (s).

Dividing this equation by the maximal life span generalizes the equation to different maximal life spans and transforms times in modes into proportions of life span

$$LS_p = 1 - A'T_{sp} \quad (1)$$

where LS_p = life span, T_{sp} = proportion of life span spent in search. Now A' is the amount by which life span is reduced when the adult butterfly takes in no nectar.

Next, the proportional number of resources discovered during a lifetime is the sum of the relative success of the two modes multiplied by the number of resources. First, the approximate number of resources located is represented as

$$\text{Area} \sim (1 - T_{sp}) + (T_{sp}B) \quad (2)$$

where Area = the area searched and B = the relative efficiency of search, or similarly stated, the relative amount of area covered by exclusive search compared to nonexclusive search.

Resources are assumed to be uniformly spaced and so the number of resources located is the product of the area searched (Area) and the density of resources. For replicating finite resources, resources change through time using a linear function and so the density of resources across a lifetime is the integral of the resource amount function multiplied by equation (2). As either death of the butterfly or the total extinguishing of resources may come first then the equations given below follow:

$$294 \quad LS_p < \frac{1}{c} \rightarrow \int_0^{LS_p} 1 - ct \, dt = LS_p - c \frac{LS_p^2}{2} \quad (3a)$$

$$295 \quad \frac{1}{c} < LS_p \rightarrow \int_0^{\frac{1}{c}} 1 - ct \, dt = \frac{1}{c} - \frac{1}{2c} \quad (3b)$$

296 where c = the rate of diminishment of resources over time and $1/c$ the point at which resource
297 densities are 0.

298 Multiplying (3a) and (3b) by (2) produces equations for the relative number of resources,
299 N_{items} , located during a lifetime, which can be evaluated in response to A' , B and c , the cost of
300 exclusive search, its relative effectiveness and the rate of resource diminishment,
301 respectively.

$$302 \quad LS_p < \frac{1}{c} \rightarrow N_{\text{items}} = 1 - A'T_{sp} - c \frac{(1 - A'T_{sp})^2}{2} ((1 - T_{sp}) + T_{sp}B) \quad (4a)$$

$$303 \quad \frac{1}{c} < LS_p \rightarrow N_{\text{items}} = (\frac{1}{c} - \frac{1}{2c}) ((1 - T_{sp}) + T_{sp}B) \quad (4b)$$

304 For the analysis A' was estimated at 0.5 for *M. jurtina*, predicting a loss of 50% of life span
305 when consuming no resources (Evans, Sibly, et al., 2019b; Lebeau et al., 2016a; Vande Velde
306 & Van Dyck, 2013). The effects of variables B and c were then evaluated to determine how
307 much exclusive search individuals should undertake to maximize the number of resources
308 encountered. Presented results include evaluating T_{sp} after conditioning on c for values of 0
309 and 1, relating to conditions where resources do not diminish (females locating host plants),
310 and when resources are at 0 at 100% of total life span (assuming male life span is adapted to
311 available reproductive opportunities).

312

313 <H1>Results

314 <H2>Cluster analysis across habitat types and sexes

Four clusters were selected by silhouettes of the time budget data, and *K*-means clustering with four centres explained 80.8% of the variance in time budgets. The probability of an individual being included in a particular cluster grouping was strongly predicted by sex and habitat type (Table 1). Clusters consisted of four main groupings: (1) a combination of basking, inactivity and flying that was shared in equal proportions between the sexes; (2) large amounts of time in flight, the majority male, and located predominantly in nectar-poor habitats; (3) inactivity, which was mainly female and split equally between habitat types; and (4) nectaring, containing a higher proportion of females and found exclusively in nectar-rich habitats. For convenience, we use the descriptive labels ‘high-bask’, ‘high-flight’, ‘high-inactive’ and ‘high-nectar’, respectively.

Males had an increased probability of classification in the ‘high-flight’ cluster and a reduced probability of classification in the ‘high-nectar’ cluster compared to females. In nectar-rich habitats, ‘high-flight’ was reduced relative to ‘high-inactive’, meaning ‘high-flight’ was more likely in nectar-poor habitats. The probability of being grouped into ‘high-bask’ also increased in nectar-rich habitats. In contrast, there was no significant effect of habitat on the probability of classification of ‘high-nectar’ relative to ‘high-inactive’. This may appear counterintuitive, but it is due to the baseline condition of ‘high-inactive’ being common in nectar-rich patches (Fig. 1).

<H2>Clusters within sexes and habitat types

When data were split by sex and habitat, different clusters were identified, with three groups for females in nectar-rich habitat: (1) ‘high-nectar’; (2) ‘high-inactive’; and (3) ‘high-bask’ a group that also contained inactivity, flying and oviposition. Male behaviour was more diversified with five groups identified: (4) ‘high-flight’; (5) ‘high-nectar’; (6) ‘high-inactive’;

(7) ‘high-bask’ a group similar to that identified in females that contained basking and small amounts of the other behaviours; and (8) an additional group containing a majority of switches between inactivity and flight, termed here ‘mixture’. Males and females had similar behavioural groupings in nectar-poor habitats, consisting of either ‘high-flight’ or ‘high-inactivity’. For both sexes in nectar-rich habitats, insolation strongly influenced cluster allocation (Table 2). In nectar-poor habitat, behaviour for both sexes was less diverse with only two cluster groupings identified corresponding with a ‘high-flight’ and ‘high-inactive’ group.

<H2>Movement behaviour in sex-specific clusters

In nectar-rich habitats, step distances were found to differ between the clusters for both males ($F_4=7.86$, $P<0.001$) and females ($F_2=4.97$, $P<0.001$). For males, Tukey’s test showed differences mostly between ‘high-nectar’ and the other groups (‘high-inactive’, ‘mixed’ and ‘high-flight’) with steps shorter for ‘high-nectar’ (Fig. 2a). ‘High-flight’ and ‘mixed’ were also significantly different. Males’ turning angles varied between ‘high-nectar’ and all other groups, with turning angles more tortuous in ‘high-nectar’ (Fig. 2b). There were also differences between ‘high-flight’ and ‘high-bask’, and ‘high-bask’ and ‘mixed’ (full results Table S3 and S4). For females, step distances were only significantly different between ‘high-nectar’ and ‘high-inactive’ (Fig. 2c). There were also differences in turning angle between ‘high-nectar’ and all other groups (Fig. 2d).

<H2>Consequences of behaviour modes for movement rates

The individual-based random-walk model was used to evaluate the effect of implementation of behavioural differences on predicted movement rates within nectar-rich habitats. For both

males and females, the inclusion of modes and intraspecific variation had a large effect on the mean and shape of the resultant displacement distribution (Fig. 3). Overall, the mode and intraspecific models produced longer-tailed distributions than the pooled model, although the models differed between the sexes. For males, the mean displacement of the mode model was lowest (201 ± 1 m) followed by the intraspecific (239 ± 3 m) and then the pooled model (302 ± 1 m; Fig. 3a). However, the reverse was the case for the females where the mode model had the largest mean displacement (106 ± 1 m), the intraspecific model was intermediate (101 ± 2 m) and the pooled model the lowest (70 ± 1 m; Fig. 3b).

<H2>Modelling optimal time in exclusive search

The optimal amount of time in exclusive search mode (T_{sp}) is shown against the efficiency of exclusive search (B) and the rate of resource diminishment (c) in Fig. 4a. In general, when efficiency is low and resources only slowly diminish, butterflies should spend little or no time in exclusive search and should locate sex-specific resources as a by-product of behaviour that maximizes survival. However, if resources diminish very quickly and exclusive search is efficient, butterflies should spend all their time in exclusive search with a subsequent sacrifice of life span. With no resource diminishment (Fig. 4b), a case likely to be representative of *M. jurtina* which feeds on common grasses, females should only spend time in exclusive search when its efficiency is 1.5 times that of survival/foraging behaviour. With a higher rate of resource diminishment (Fig. 4c), a case representative of male *M. jurtina* locating receptive females in this monandrous species, butterflies should always spend some time in exclusive search even if it is only marginally more efficient than normal behaviour, increasing up to more than 75% when search is twice as efficient.

387

388 <H1>Discussion

389 In this study, we explored various aspects of the behavioural modes of the model butterfly *M.*
390 *jurtina*. In nectar-poor regions, both sexes were characterized by modes of either inactivity or
391 high amounts of time in flight. In nectar-rich habitats, however, sex-specific behaviour was
392 more diversified. Females had three modes, ‘high inactivity’, ‘high-basking’ and ‘high-
393 nectar’, and males additionally had ‘high-flight’ and ‘mixed’ modes consisting of a high
394 proportion of time in flight, and transitions between flight and inactivity. Movement
395 parameters differed between the modes (Fig. 2) and their inclusion in a random-walk model
396 had large effects on the shape of the displacement distribution (Fig. 3). Including modes also
397 had different effects on the spatial distribution of the sexes, with males moving less on
398 average compared to the other model versions while for females including modes increased
399 movement. Finally, the model of exclusive search behaviour demonstrated the general
400 conditions under which exclusive search is favoured (Fig. 4) and indicated the time that
401 should be spent in exclusive search given its cost, its effectiveness and the rate of resource
402 diminishment.

403 In nectar-poor regions, behavioural modes were similar for the sexes, and consisted of either
404 inactivity or spending a high proportion of their time in flight. As these areas have low
405 resource densities, a parsimonious explanation for these modes is that flight and inactivity are
406 the only possibilities, thus requiring no account of motivation. However, as increasing
407 movement rate in response to poor-quality habitats is common to many butterfly species
408 (Fownes & Roland, 2002; Odendaal et al., 1989; Ovaskainen, Rekola, Meyke, & Arjas, 2008;
409 Roland, Keyghobadi, & Fownes, 2000; Schtickzelle et al., 2007; Zalucki & Kitching, 1982),
410 and is a general response across many taxa (Fryxell et al., 2008; Haskell, 1997; Smith, 1974;

Zollner & Lima, 2005), the ‘high-flight’ mode observed is likely to be a specific behaviour pattern aimed at quickly moving *M. jurtina* out of poor-quality areas and not only a by-product of low resource density. That movement in these areas is also faster and straighter (Evans et al., 2020b) suggests that a ‘high-flight’ mode may correspond with exploratory or dispersive movement (Delattre et al., 2010), as seen in many other taxa (Patterson, Thomas, Wilcox, Ovaskainen, & Matthiopoulos, 2008) and is probably distinct from behaviour occurring in nectar-rich habitats. This is most notable in the females where, in nectar-rich habitats, the ‘high-flight’ mode was absent. Therefore, it seems likely that butterflies in poorer-quality regions are either unable to fly due to thermal or physiological constraints, or switching to flying frequently.

In nectar-rich habitats, males and females had different modes that largely corresponded with previous work evaluating sex-specific behaviour in butterflies (Brakefield, 1982a; Scott, 1974; Shreeve, 1992; Wiklund, 2003). Females were either inactive, basking or nectaring. This low-energy regime corresponds with maximizing adult life span which, from the exclusive search model, is an optimal strategy. Females of *M. jurtina* progressively mature eggs through their life span (Scali, 1971), probably producing a strong correspondence between fitness and survival time. Oviposition was seen in the ‘high-bask’ cluster, although still observed rarely. Therefore, it was not possible to determine a distinct oviposition mode. Females of *M. jurtina* have a flight pattern that does correspond with oviposition, flying low over the ground and laying a series of single eggs each a short distance apart, although this was indistinct from other behavioural modes probably because of the timescale of our observations. Males had two additional behaviours, ‘high-flight’ and ‘mixed’. ‘High-flight’ we consider to correspond with a behaviour termed patrolling (Brakefield, 1982a) where males fly for longer periods as they search for receptive females and engage less in behaviours such as nectaring and inactivity. The exclusive search model suggests this is an

optimal strategy for maximizing the number of receptive females located, incentivizing some sacrifice of life span. The other mode termed ‘mixed’ may relate to perching behaviour seen in the grassland species *Coenonympha pamphilus* (Wickman, 1985) and *Lasiommata megera* (Dennis, 1982), the woodland species *Pararge aegeria* (Bergman et al., 2007; Shreeve, 1984; Wiklund, 2003) and also possibly *M. jurtina* (Brakefield, 1982a). This is an alternative mate-finding strategy where males wait inactive and chase females as they pass by. However, it is also possible that, as butterflies were followed opportunistically, the ‘mixed’ mode consisted of transitions between the other modes, rather than a specific behavioural pattern; therefore, we limit our interpretation at this time.

For both sexes, movement parameters were found to differ largely between ‘high-nectar’ and the other modes, although ‘high-flight’ also differed for males (Fig. 2). The short step distances and tortuosity of ‘high-nectar’ are probably a by-product of moving from flower to flower and slower flight speeds may relate to the ability to survey potential resources in flight (Chittka, Dyer, Bock, & Dornhaus, 2003; Chittka, Skorupski, & Raine, 2009). ‘High-flight’ also probably results in longer step distances and straighter flight paths as males attempt to survey larger areas when searching for females. The variation in the movement parameters and the effect of implementing behaviour in the individual-based random-walk model combined to have large effects on displacement distributions (Fig. 3). The relationship of the mode model to the intraspecific and pooled models was qualitatively different between the sexes. We attribute this to the mode model replicating transition through behavioural states, generating females that move more than was observed individually (intraspecific model), while also switching between the more and less diffusive movement states absent in the pooled model. For males, the mode model was intermediate between the intraspecific model, where observed individual differences are extrapolated, to the pooled model where individuals are identical. Disentangling the effects of intraspecific variation versus

behavioural modes is challenging. Models that incorporate intraspecific variation in movement rate have been successful in replicating realistic movement patterns (Brown & Crone, 2016; Walters, Hassall, Telfer, Hewitt, & Palutikof, 2006) and there is good evidence that traits such as metabolic rate consistently influence interindividual variation in movement (Mattila, 2015; Ovaskainen, Smith, et al., 2008). Consequently, both movement modes and syndromes (Sih, Bell, & Johnson, 2004; Spiegel, Leu, Bull, & Sih, 2017) are likely to be important for understanding species movement rates. We suggest though, that for short timescale observations (tens of minutes), such as those for butterflies, it may be challenging to separate movement modes from individual propensities and their effect on movement predictions needs careful consideration.

The exclusive search model provides the general conditions (Fig. 4) under which exclusive search is favoured. In two examples, we considered nondiminishing and diminishing resources, representing female *M. jurtina* searching for host plants and males searching for receptive females. For females, exclusive search was only advantageous when it was more than 1.5 times more successful for locating host plants than normal behaviour. As grasses are relatively ubiquitous, females of *M. jurtina* can be predicted to have little exclusive search behaviour and no specific mode for host plant search was in evidence. Similarly, as females are monandrous, we expected males to engage in exclusive search and we found evidence of patrolling behaviour. Generally, the amount of exclusive search could relate not only to properties of flight paths but also to the distribution of resources. For example, pierid species lay eggs on brassicas (Thomas, 2010) which are rarer and more clumped than grasses, probably requiring an exclusive search for locating plants, and female pierids engage in active search for host plants (Dennis & Hardy, 2007; Root & Kareiva, 1984). Likewise, in polygamous species, or those with active females, we may assume less necessity for the males to engage in exclusive search. Thus the model provides a framework to view the sex-

specific behaviour of many butterfly species from estimated parameters such as the cost of search (widely available e.g. Lebeau et al., 2016b; Nütepöld & Boggs, 2015; Woods, Wood, Ebersole, & Stevenson, 2010), the effectiveness of search and the rate of resource diminishment. Some factors are not taken into account by the model, such as the effect of different tactics used by perching or lekking butterflies (Alcock, 1985; Brown & Alcock, 1990; Scott, 1974), which would strongly influence both the success and the cost of search (Dennis & Shreeve, 1988), although the model could be adapted through the appropriate parameterization. Further, the assumption of uniform resources over the landscape is simplistic, and it would be useful to evaluate how changes in the efficiency of search over time might influence the use of exclusive search.

A limitation of this study is that the description of modes is related to the duration of observations, risking the methods influencing our interpretation of the results. We feel here though that the ability to group behaviour into meaningful clusters that correspond with previous observations of butterfly behaviours demonstrates timescales at tens of minutes are appropriate. Further, an ability to separate intraspecific variation from behavioural modes would be enhanced by following butterflies for longer periods and attempting to observe switches between modes within individuals. This is feasible, but due to the intensive nature of the data collection would be time consuming to accumulate for a large sample of individuals. Finally, the main focus of the study was behaviour operating within nectar-rich habitats, and extrapolating movement to complex habitats will require a better understanding of how modes change in response to habitat types, varying resource densities, habitat edges and an individual's age and condition (Conradt, Bodsworth, Roper, & Thomas, 2000; Conradt & Roper, 2006; Delattre et al., 2010; Kallioniemi, Zannese, Tinker, & Franco, 2014; Mair, Thomas, Franco, & Hill, 2015; Polic, Fiedler, Nell, & Grill, 2014; Schneider, 2003). These other factors may be particularly important for sedentary species like *M. jurtina* for which

mark–recapture studies find lower mean dispersal estimates (45–414 m; Schneider, Dover, & Fry, 2003) than would be expected from direct extrapolations of movement observations.

In conclusion, we have identified the importance of behavioural modes for the fitness and movement behaviour of the model species *M. jurtina*. Our results provide two main innovations. First, we evaluated, in a movement model, the effect of behavioural modes on predicted movement rates. Second, we produced a search model that conceptualized the trade-off between searching for the resources necessary for reproductive fitness and searching for those for sustaining life span. The balance between reproduction and survival is central to the life history of all species and we hope that our search model, targeted at understanding this trade-off in *M. jurtina*, will provide a useful route to evaluating how butterflies and other species maximize their fitness given the resources they utilize and the constraints acting on their mobility and perception. Our movement model demonstrated that the different methods of incorporating interindividual variability have large effects on movement predictions. In particular, we identified the challenge of disentangling intraspecific variation from context-specific behavioural modes. Further work attempting to evaluate butterfly movement in light of these concepts is likely to allow better integration of the wealth of behavioural information on butterflies when investigating aspects of their movement ecology such as habitat use, optimal foraging and dispersal.

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854 **Table 1.** Coefficients (\pm SEs) from the multinomial regression

	Intercept	Sex (male)	Habitat (nectar- rich)	Air temperature
3 (High-inactive)	-	-	-	-
1 (High-bask)	-4.85 (1.29)***	-0.33 (0.36)	1.93 (0.76)*	0.08 (0.05)
2 (High-flight)	-0.42 (1.03)	3.48 (0.49)***	-2.24 (0.38)***	-0.07 (0.05)
4 (High-nectar)	-7.86 (23.79)	-0.79 (0.33)*	9.91 (23.77)	-0.12 (0.05)*

855 Coefficients show changes in log odd ratios of a time budget occurring in a cluster relative to
856 the baseline cluster for a unit change in the predictors. The baseline here is the inactive group
857 ('High-inactive'). Significance scores for coefficient estimates were produced using a two-
858 tailed Wald test.

859 * $P < 0.05$; *** $P < 0.001$.

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861

Table 2. Coefficients from the multinomial regression of clustering within nectar-rich habitats

	Female		Male	
	Intercept	Insolation	Intercept	Insolation
High-inactive	-	-	-	-
High-bask	-1.01***	2.17×10^{-6}	-1.67***	2.43×10^{-5}
High-nectar	-1.5***	9.46×10^{-6} ***	-3.67***	2.28×10^{-5} ***
High-flight †	--	--	-3.8***	2.56×10^{-5} ***
Mixed †	--	--	-1.93***	1.02×10^{-5} ***

Coefficients show changes in log odd ratios of a time budget occurring in a cluster relative to the baseline cluster for a unit change in the predictors. The baseline here is the inactive group ('High-inactive'). Significance scores for coefficient estimates were produced using a two-tailed Wald test.

*** $P < 0.001$.

† Cluster only observed in males.

Figure 1. The average duration of behaviours within clusters and the proportion of individuals grouped in clusters across sex and habitat types. Left-hand panels (a-d) show cluster groups with bars representing the mean proportions of time the behaviour was performed in the cluster, middle panels show the proportion of the different sexes grouped in the cluster, and the right-hand panels show the proportion of the habitat types (nectar-poor and nectar-rich) in which the cluster was observed. ‘Ovi’ refers to oviposition an activity rare across all clusters.

Figure 2. Movement parameters in selected cluster groups. (a) Step distances and (b) turning angles for males in the ‘high-nectar’ versus ‘high-flight’ clusters. (c) Step distances and (d) turning angles for females in the ‘high-nectar’ versus ‘high-inactive’ clusters. Pairings were chosen as examples where both step distances and turning angles were significantly different between the groups.

Figure 3. Comparison of displacement predictions from the random-walk models. (a) Males and (b) females.

Figure 4. (a) The optimal amount of time butterflies should spend in exclusive search T_{sp} , given the rate of resource diminishment c and the relative search effectiveness of exclusive search over normal behaviour B . (b) The optimal time when $c = 0$ and (c) the optimal time when $c = 1$ representing no resource diminishment and total resource diminishment at maximal life span, respectively.