Distance sampling and the challenge of monitoring butterfly populations

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Summary

1. Abundance indices generated by the UK Butterfly Monitoring Scheme (UKBMS) have been influential in informing our understanding of environmental change and highlighting UK conservation priorities. Here, we critically evaluate the standard 'Pollard Walk' methodology employed by the UKBMS.

2. We consider the systematic sampling biases among different butterfly species and biotopes using distance sampling. We collected over 5000 observations on 17 species using distance sampling at 13 study sites in England and Wales. We fitted detection functions to explore variation in detectability among species and sites.

3. Our results suggest that around one-third of individual butterflies in the Pollard Walk box were missed. However, detectability varies markedly among species and sites. We provide the first species-specific estimates of detectability for converting Pollard Walk data into population densities. A few species show no drop off in detectability and most require only a modest correction factor, but for the least detectable species, we estimate that 3/4 of individuals are not recorded.

4. Much of the variation among sites is explained by substantially higher detectability among sites in England than in Wales, which had different recorders. Biological traits have only limited explanatory power in distinguishing detectable vs undetectable species.

5. The variation in detectability is small compared with the variation in true abundance, such that population density estimates from the Pollard Walk are highly correlated with those derived from distance sampling.

6. These results are used to evaluate the robustness of the Pollard Walk for comparisons of abundance across species, across sites and over time. UKBMS data provide a good reflection of relative abundance for most species and of large-scale trends in abundance. We also consider the practicalities of applying distance sampling to butterfly monitoring in general. Distance sampling is a valuable tool for quantifying bias and imprecision and has a role in surveying species of conservation concern, but is not viable as a wholesale replacement for simpler methods for large-scale monitoring of multispecies butterfly communities by volunteer recorders.

Key-words: apparency, bias, butterfly monitoring scheme, detectability, effective strip width, habitat, mixed models, Pollard Walk, sampling, transect

Introduction

Population abundance is a critical variable in ecology (McGill 2006): abundance data are required to understand the basic

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population dynamics of species, as well as to provide information on the state of biodiversity (Loh *et al.* 2005). One of the largest datasets on non-pest insect population dynamics comes from the UK Butterfly Monitoring Scheme (UKBMS, Pollard & Yates 1993; Rothery & Roy 2001; Fox *et al.* 2006). The UKBMS has provided data on the abundance of butterfly populations for over three decades, and over 850 sites are now monitored annually (Botham *et al.* 2008). The methods developed for the UKBMS have been adopted by monitoring schemes in several other countries (van Swaay *et al.* 2008). Data from the UKBMS have provided valuable insights into the population-level effects of land-use and climate change (e.g. Roy & Sparks 2000; Roy *et al.* 2001; Warren *et al.* 2001; Brereton *et al.* 2007; Oliver *et al.* 2009; Isaac *et al.* 2011). These findings, allied with certain aspects of butterfly biology (rapid life cycle, microhabitat requirements), make butterflies a key indicator of environmental change (Thomas *et al.* 2004; Thomas 2005).

The majority of UKBMS data are collected using a fixed-width transect count method, in which recorders count individual adult butterflies along set routes that are subdivided into sections (Pollard *et al.* 1975; Pollard 1977; Pollard & Yates 1993). The method is known as the butterfly transect method or Pollard Walk: we use the latter to distinguish it from other transect-based methods. A key feature of the Pollard Walk is the imaginary moving box of 5 m each side (250 cm on both sides of the transect line): individuals observed within the box are counted, whilst those outside are ignored. The method allows large quantities of data to be collected on butterfly communities, using simple rules that can be learned and adopted quickly. Intensive field studies have shown that counts from Pollard walks are closely correlated with absolute numbers of butterflies present, when the survey design representatively samples the site (e.g. Pollard 1977; Thomas 1983, 1991; Warren, Thomas, & Thomas 1984; Warren, Pollard, & Bibby 1986; Sutcliffe, Thomas, & Moss 1996; Haddad *et al.* 2008), but see Harker & Shreeve (2008).

Analyses of Pollard Walk abundance estimates generally assume that a constant proportion of the butterfly population is surveyed. This may be reasonable when comparing the same site year on year for a particular species but may not hold true for comparisons between biotopes (e.g. Brown & Boyce 1998; Haddad *et al.* 2008). In addition, vegetation changes associated with climate change or management regimes may affect the detectability of butterflies and generate apparent changes in abundance or mask real trends (Davies *et al.* 2006; Dennis & Sparks 2006). Systematic changes in detectability over time would further reduce the degree to which abundance estimates are comparable, thus making it difficult to draw either theoretical or applied conclusions from such data. Moreover, potential differences exist in the visibility of different species (e.g. Thomas 1983; Pollard & Yates 1993). Dennis *et al.* (2006) found that visual apparency of British butterflies at a national scale is correlated with size, wing colour and flight behaviour. For this reason, there have been few attempts to use UKBMS data for interspecific comparisons of abundance (Cowley *et al.* 2001; Isaac *et al.* 2011). However, no methodological assessment has

been conducted on the relative detectability of butterfly species at the biotope level.

Accurate population estimates with defined precision are increasingly being demanded in relation to the conservation of rare species and analyses of population viability and metapopulation dynamics. This is particularly true for species with low or fluctuating abundance and patchy or restricted distribution (Brown & Boyce 1998; Boughton 2000; Powell, Busby, & Kindscher 2007). One problem with the Pollard Walk is that it does not generate confidence intervals around individual estimates of abundance, so the precision of UKBMS data is unknown (see also Haddad *et al.* 2008). Thus, a critical evaluation of the bias and precision of the Pollard Walk is important for both fundamental and applied ecological questions.

The ecological literature contains many techniques for estimating absolute population size (e.g. Southwood & Henderson 2000). Among the most widely used is distance sampling (Buckland *et al.* 2001; Thomas *et al.* 2010), which has been shown to give accurate and unbiased estimates of population density when not all individuals within a surveyed area are sampled. Distance sampling is usually transect based but can also be applied to point counts. It works by fitting a detection function to observations at known distances. The shape of this function defines the effective strip width (ESW), which provides a simple measure of detectability. ESW is the distance at which the number of individuals observed further away is estimated to equal the number of individuals closer to the line that were missed. Population density can be calculated as the number of individuals counted divided by [ESW*2*distance travelled]. The published literature contains few applications of distance sampling to butterflies (Brown & Boyce 1998) and none in the context of validating monitoring data (see Newson *et al.* 2008 for an application to bird monitoring).

The key challenge we address here is the extent to which the relative abundance estimates derived from the Pollard Walk are comparable among species and among sites. We use distance sampling to estimate the variation in detectability of butterflies on UKBMS transects and compare abundance estimates from the two methods. We explore the limitations of the Pollard Walk and address the potential for distance sampling as a tool in monitoring butterfly populations. Our inferences are based on estimates of the detection function within the Pollard Walk box on existing UKBMS transect routes, which do not represent a random sample of the landscape. We do not address the issue of survey design, which is paramount for obtaining unbiased estimates of animal abundance (Thomas *et al.* 2010). Our primary focus is on how distance sampling can inform the interpretation of data collected on existing UKBMS routes.

We address four specific research questions. First, for each species, what proportion of butterflies is missed by the Pollard Walk? Second, what is the magnitude of variation in detectability among species and sites? Third, to what extent is detectability explained by butterfly biology and biotope characteristics? Finally, how well correlated are Pollard Walk and distance-based estimates of population density?

Our results have potentially wide-reaching implications for butterfly monitoring. Converting Pollard Walk data into absolute abundances would greatly enhance the value of the data already collected, providing new opportunities for analyses of the viability of populations, and make the data amenable to studies of community ecology and macroecology. This is a great opportunity, both to enhance studies of past population changes and to increase the rigour of future monitoring in Europe and elsewhere (Haddad *et al.* 2008; Nowicki *et al.* 2008). Moreover, understanding detectability will inform new techniques for monitoring rare species of particular conservation concern, and in the wider countryside (Thomas 1983; Roy, Rothery, & Brereton 2007; Nowicki *et al.* 2008; van Swaay *et al.* 2008).

Materials and methods

DATA COLLECTION

Fieldwork was carried out at nine sites in north Wales and four in southern England (Table 1). Welsh field sites fall within an area of 35 km^2 on the Creuddyn Peninsula and Anglesey and cover a range of biotopes (Cowley *et al.* 2001). Transects were laid out to sample representative habitats (following Pollard *et al.* 1975). Some transects followed existing footpaths, whilst others traversed open land. English sites were all UKBMS transect routes on south-facing chalk grassland slopes with varying degrees of scrub invasion and grazing pressure: two are situated on the Dunstable Downs in Bedfordshire and two on the North Downs in Surrey.

The Welsh study was conducted by D.M. Shuker between 18th May and 22nd September 1998 ($n = 2256$ butterflies recorded). The English study was conducted by A.M. Weddle between 28th June and 25th July 2006 ($n = 3304$). Perpendicular distances were estimated by eye to the nearest 10 cm in the Welsh sites (0, 5, 15, 25 etc) and to

the nearest 5 cm at English sites. Both studies followed the UKBMS criteria for weather and time of day (Pollard & Yates 1993). Butterflies were identified to species level, with the exception of Small and Green-veined whites (*Pieris rapae* and *P. napi*) and, in the English study, Small and Essex skippers (*Thymelicus sylvestris* and *T. lineola*), which could not be distinguished reliably in flight.

ANALYTICAL APPROACH

We expect that detectability varies systematically among species and sites. The nature of this variation is of primary interest (Question 2), but also means that neither the raw observations nor the derived strip widths can be considered mutually independent. For these reasons, we fitted separate detection functions to each site–species combination, pooling the data across visits, and analysed the resulting strip widths using linear mixed-effects models. This provides an effective means for partitioning and estimating the variance in detectability but is not optimal for robustly estimating population density (Thomas *et al.* 2010). Our measures of population density (Question 4) should therefore be treated with caution. To test whether this approach compromised our conclusions, we repeated our analysis on the factors associated with detectability (Question 3) using the multiple covariate distance sampling (MCDS) engine (Thomas *et al.* 2010). These results are presented in the Supporting Information.

EFFECTIVE STRIP WIDTHS

We estimated ESWs using Distance v6·0 (Thomas *et al.* 2010). We stratified our analyses by each unique combination of species and study site, after removing all combinations with fewer than 20 observations. Although this is smaller than the recommended minimum sample size of 60 (Thomas *et al.* 2010), we feel justified in using a smaller number because our aim is to explore variation in detectability not the precise estimation of population density. This restricted data set consists of 5363 observations on 17 species (50 site-by-species combinations, Table 2). Preliminary analysis revealed that certain

Table 1. Description of study sites. *L* is the transect length (in metres), *N* is the number of walks, and VH is the index of vegetation height

Location	Coordinates	Name	Description	L	N	VH
Dulas Valley, Conwy	53°16′49″N 3°38′25″W	DV1	Lots of low *Cotoneaster*, some scrub and grasses, quite open	306	18	2
		DV2	Mix of longer grasses and open turf, some scrubby vegetation	255	11	4
Glan Conwy, Conwy	53°16′33″N 3°47′51″W	GC	RSPB reserve. Open, grasses and herbs, tall in places	640	11	3
Great Orme, Llandudno	53°19′45″N 3°51′12″W	GOI	Mix of *Rubus* scrub and grassland, generally quite scrubby	160	14	4
		GO2	Short, close cropped turf, very open	445	5	1
		GO3	Short, close cropped turf, open with scrub	515	3	1
Newborough Warren, Anglesey	53°10′37″N 4°22′40″W	NW1	Taller grasses and herbs	1375	3	5
		NW2	Mix of open turf and longer grasses	515	3	3
Llangwstenin, Conwy	53°17′46″N 3°46′17″W	LST	Quite tall woody scrub	330	3	6
Bison Hill, Dunstable	51°51′44″N 0°32′45″W	BH	SSSI. Thick grass. Ungrazed, mown annually. Sward height ~70 cm	620	5	3
Whipsnade Zoo, Dunstable	51°51′07″N 0°33′05″W	WZ	Heavily grazed by wallabies and Chinese water deer. Sward height < 5 cm	1450	5	1
Pewley Downs, Guilford	51°13′48″N 0°33′24″W	PD	Grass with some scrub invasion. Ungrazed, but mown annually. Sward height ~75 cm	630	5	4
Denbies Landbarn, Dorking	51°14′17″N 0°22′35″W	DL	Grazed by ponies. Sward height ~35 cm	740	5	2

Table 2. Number of butterflies recorded for each species–site combination. Combinations with fewer than 20 observations were excluded. Site names as in Table 1. *The Essex skipper does not occur in North Wales

	Welsh sites									English sites			
	DV1	DV2	GC	GO1	GO2	GO3	NW1	NW2	LST	BH	WZ	PD	DL
Pieridae													
Brimstone												55	
Large White										28		29	
Small/Green-veined White										24		42	
Lycaenidae													
Brown Argus	46	32		53									
Chalkhill Blue												390	352
Common Blue			36										
Silver-studded Blue	47	62		399									
Nymphalidae													
Gatekeeper	50	38	216	58						75		78	27
Grayling					71	190							
Marbled White										260		96	185
Meadow Brown		38	299	96			22	20		254	50	345	225
Ringlet							63	102		154		66	23
Small Heath		28											137
Speckled Wood									24				
Hesperiidae													
Dingy Skipper	62	71											
Large Skipper												30	
Small/Essex Skipper			67*							185		35	28

combinations contained a high proportion of observations on butterflies basking on the transect path, thus violating the assumption that organisms are randomly positioned with respect to the transect line (Thomas *et al.* 2010). This phenomenon tends to give distance distributions that are strongly spiked at zero, resulting in underestimated detection functions. To circumvent this problem, we analysed grouped data, selecting an initial bin width broad enough to remove the apparent spike (Buckland *et al.* 2001) and, as there were no other heaping problems, simply using this width throughout to give ten equally spaced distance bins. In practice, binning the data in this way had little effect on the estimated ESW for most combinations (Table S1), and the Pearson correlation between ESWs using binned and raw distances was 0·95. Similar estimates were produced using differing numbers of bins (Fig. S1).

For each combination, we sought the best description of the detection function by fitting the six models suggested by Thomas *et al.* (2010: uniform plus cosine/polynomial adjustments, half-normal plus cosine/hermite polynomial adjustments, hazard rate plus cosine/polynomial adjustments) and selecting models in terms of goodness-of-fit statistics and AIC (Akaike Information Criterion), following visual inspection of the data. Distance sampling data are generally truncated at some specified distance to reduce the influence of outliers (Thomas *et al.* 2010). We generated two sets of ESWs using different truncations: one truncated at the 95th distance percentile for each combination (following Thomas *et al.* 2010) and one with a universal truncation distance of 250 cm from the transect line (to give the width of the standard Pollard Walk box: 37% of observations were made at > 250 cm). The full set of ESWs is presented in the Supporting Information (Table S1). We used the 250-cm truncation to estimate species-specific correction factors for the UKBMS (Question 1). We used both sets of data to explore the variation in detectability (Question 2), the factors explaining detectability (Question 3) and compare estimates of population density (Question 4).

STATISTICAL MODELLING

We used linear mixed-effect models to partition the variance in ESW between sites and species and to test hypotheses about detectability. All analyses were conducted using the *lme4* package (Bates, Maechler, & Dai 2008) in *R* (R Development Core Team, 2008). We weighted each of the 50 ESWs by the square root of the number of observations inside the truncation distance, rescaled to have a mean of 1. Weighting the data in this manner reduces the impact of combinations with small sample sizes, where ESW is likely to have been estimated imprecisely. Visual inspection of the residual distribution indicated that input variables did not require transformation, although each variable was centred on zero for modelling.

We first estimated species-specific ESWs using a model with species as a fixed effect. These values were converted into correction factors (Question 1) by dividing them into the common truncation distance of 250 cm. We then estimated the variance components (Question 2) by fitting models with random effects for Site and Species and no fixed effects. Finally, we tested a suite of hypotheses about differences in detectability among species and sites (Question 3). We used two site traits and three species traits to test these hypotheses. The site traits were study (England vs Wales) and vegetation height measured from 1 (short grass) to 6 (high scrub, see Table 1). Species traits were wingspan (in mm), bask mode (dorsal vs lateral) and colour measured from 1 (dull) to 5 (very bright), all using data presented in Dennis *et al.* (2006). We modelled vegetation height and colour as continuous variables. We fitted all main effects and first-order interaction terms, and then sequentially removed non-significant terms to arrive at a minimum adequate model (MAM). Significance of fixed effects was estimated by sampling 10 000 times from the posterior distribution of the fitted parameters using Markov Chain Monte Carlo methods (Bates, Maechler, & Dai 2008).

BUTTERFLY POPULATION DENSITY

We made three estimates of butterfly population density (ha^{-1}) for each site–species combination (Question 4), using (i) Pollard Walk data (i.e. assuming no variation in detectability), (ii) distance sampling based on the 250-cm truncation and (iii) distance sampling based on the 95% truncation. We did not calculate confidence limits on the density estimates derived from distance sampling because our data were collected on a single transect at each site, thereby making it impossible to estimate variation in the encounter rate (Thomas *et al.* 2010). In addition, several combinations showed no measurable drop off in detectability: ESW for these combinations is estimated to equal the truncation distance with zero error, in spite of the small sample sizes involved (Table S1).

Results

Detection distances ranged from 0 to 1430 cm, with a median of 182 and a mean of 223 cm (Fig. 1). Across the 50 site–species combinations, the median ESW is 300 cm for the 95% truncation and 164 cm for the 250cm truncation (see Supporting Information for the full set of ESWs). These data suggest that $1–164/250 \approx 1/3$ of all individuals within the Pollard Walk box were missed.

Species-level ESWs (Fig. 2) range from under 60 cm up to the truncation distance of 250 cm and fall into three clear groups. One group consists four highly detectable species (Brimstone, Large White and Large Skipper and Small/Essex Skippers) for whom little or no correction factor is needed (i.e. the distance model indicates effectively no measurable drop off in detectability within 250 cm). Another group contains two species (Dingy Skipper and Brown Argus) with extremely short ESWs, suggesting that only around 25% of individuals are detected. The remaining 11 species show a moderate drop off in detection (135 cm < ESW < 210 cm) and for whom a modest correction factor (1·2–1·9) would be appropriate (Table 3). For nine of these intermediate species, the estimated ESW is significantly shorter than the Pollard Walk truncation of 250 cm (Fig. 2).

Despite these differences, species identity contributes only a small portion of the variance in detectability *within* the Pollard Walk box. Just 7% of the variance is among species, compared with 29% among sites and 65% residual error. However, the picture is quite different when observations beyond 250 cm are considered (i.e. using the 95% distance truncation): variance among species in detectability contributes 52% of the total, with 35% among sites and 14% owing to residual error. This difference between our two sets of ESWs reflects the fact that strip widths cannot be larger than the truncation distance, and that some species with large ESWs (notably the Large White) have few observations within 250 cm of the transect line (and therefore low weight). The total variance among the 50 ESWs is six times greater using the 95% set than using the 250-cm truncation.

The minimum adequate models of detectability (Table 4) reveal that much of the variation among sites is attributable to study: ESWs in the Welsh study were much shorter than for sites in England. Other correlates depend on the choice of truncation distance used. Within the Pollard Walk box, the only other significant correlate of detectability is the interaction between study and wingspan: each millimetre increase in butterfly wingspan leads to an increase in ESW of around 4 cm in Wales but had no significant effect among English sites. Using the 95% truncation, we find that colourful species are easier to detect: the fitted difference in ESW between the dullest species (colour = 1) and the brightest (colour = 5) is nearly three metres. We found small but non-significant positive relationships between size and detectability ($P \sim 0.07$) and the interaction between colour and wingspan ($P \sim 0.06$): each millimetre increase in butterfly wingspan leads to an increase of 8·5 cm in ESW for the brightest species but no increase for dull species. Detectability does not differ between species that bask dorsally
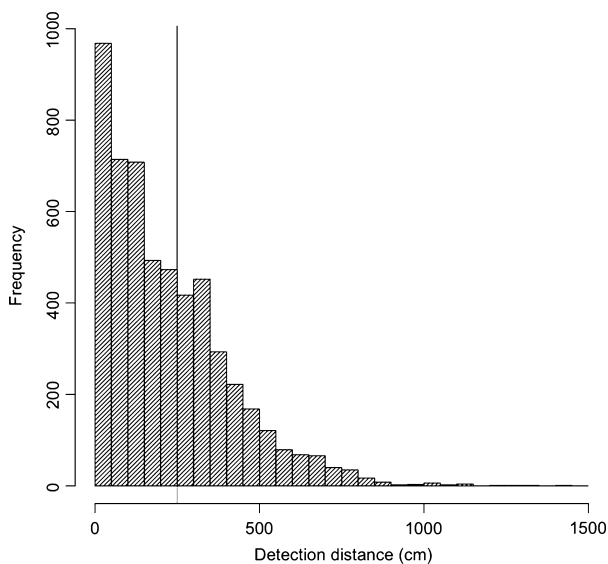

Fig. 1. Histogram of detection distances among 5363 observations of butterflies on transects. The vertical bar indicates the edge of the Pollard Walk box, outside which butterflies are not counted.
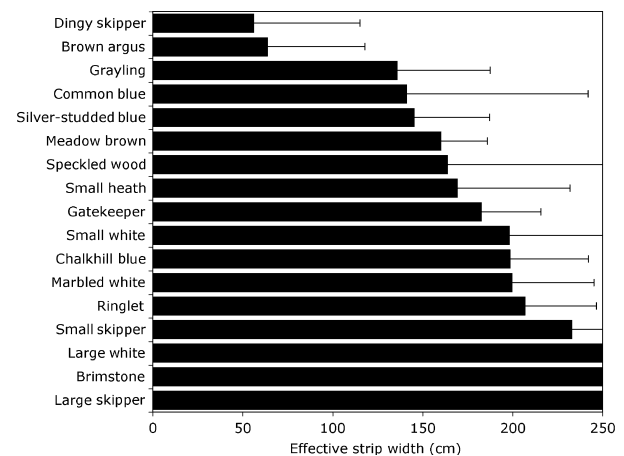

Fig. 2. Species-level strip widths (in cm) for data collected within the 250 cm Pollard Walk box. Data are parameter estimates from a model of 50 site–species combinations with species as an explanatory variable. Error bars define the 95% confidence limits.

Table 3. Species-specific effective strip widths (ESWs), associated standard errors (SE) and correction factors (CF) for the 250-cm truncation. Figures are fitted values from a linear mixed-effects model (see text for further details)

Family	Common name	Latin name	ESW/cm	SE	CF
Pieridae	Brimstone	*Gonepteryx rhamni*	250·0	51·9	1
	Large White	*Pieris brassicae*	250·0	53·0	1
	Small/Green-veined White	*Pieris sp.*	198·2	46·2	1·26
Lycaenidae	Brown argus	*Aricia agestis*	63·7	27·6	3·92
	Chalkhill Blue	*Polyommatus coridon*	198·6	22·2	1·26
	Common Blue	*Polyommatus icarus*	141·1	51·4	1·77
	Silver-studded Blue	*Plebejus argus*	145·3	21·4	1·72
Nymphalidae	Gatekeeper	*Pyronia tithonus*	182·6	16·9	1·37
	Grayling	*Hipparchia semele*	135·8	26·4	1·84
	Marbled White	*Melanargia galathea*	199·7	23·2	1·25
	Meadow brown	*Maniola jurtina*	160·2	13·2	1·56
	Ringlet	*Aphantopus hyperantus*	206·9	20·3	1·21
	Small Heath	*Coenonympha pamphilus*	169·3	31·9	1·48
	Speckled Wood	*Pararge aegeria*	163·7	56·9	1·53
Hesperiidae	Dingy skipper	*Erynnis tages*	56·2	30·1	4·45
	Large Skipper	*Ochlodes sylvanus*	250·0	55·4	1
	Small/Essex skipper	*Thymelicus sp.*	232·8	22·1	1·07

Table 4. Parameters from the minimum adequate model of the variability in detectability among species and sites (*n* = 50 combinations). *P*-values were estimated by sampling 10 000 times from the posterior distribution of the fitted parameters using Markov Chain Monte Carlo methods

	95% Truncation			250-cm Truncation		
	Estimate	SE	*P*	Estimate	SE	*P*
Intercept	428·8	30·2	< 0·0001	209·8	10·9	< 0·0001
Study (wales)	−172·8	33·2	< 0·0001	−66·2	14·5	< 0·0001
Colour	73·4	19·2	0·0002			NS
Wingspan		~0·07		−1·14	1·12	0·28
Study: Wingspan		NS		3·20	1·46	0·036

vs. those basking laterally, nor does it correlate with our index of vegetation height. Broadly similar results were obtained using the MCDS engine (Table S2), which suggests the 'proportion missed' within the Pollard Walk box is 33% in Welsh sites, compared with just 11% in English ones.

In spite of the differences in detectability we have observed, the population density estimates derived from distance sampling and the Pollard Walk are broadly similar (Fig. 3). The Pollard Walk densities tend to be underestimates because they do not take into account any drop off in detectability. Pollard Walk densities are more tightly correlated with density estimated from the 95% truncation ($r^2 = 0.933$) than from the 250-cm truncation ($r^2 = 0.789$). This is surprising because of the higher variance in ESW for the 95% truncation and because the densities based on Pollard Walk and the 250-cm truncation use the same numerator (number of butterflies) in the density estimate. Although the overall correlation is high, the degree of underestimation is extreme in a minority of cases: around 10% are underestimated by a factor of 3 or worse (dotted line on Fig. 3). The relationship between Pollard Walk and distance estimates of density is somewhat triangular: the mean

discrepancy between the two estimates is greater at high density. Naïve interpretation might suggest that populations occurring at low density tend on average to be more detectable, and that the Pollard Walk is therefore less biased for rare than common populations. However, this phenomenon is almost certainly an artefact of excluding combinations with small sample sizes: low-density populations that are difficult to detect would not generate enough data to be considered, whereas high-density populations with similarly low detectability would show up as poorly estimated by the Pollard Walk.

Discussion

Our results reveal that a sizeable proportion of butterflies are missed by the Pollard Walk and that detectability (the proportion missed) varies substantially among species and sites. Whilst previous studies have reported variation in the detectability of butterflies among species (Kery & Plattner 2007) and biotopes (Brown & Boyce 1998), ours is the first to quantify, compare and model them.

Overall, the variability among species in detectability is large. However, most of this disappears if observations outside the Pollard Walk box are excluded. This means that UKBMS data provide a good reflection of relative abundance for most species. Our species-specific correction factors (Table 3) estimate the degree to which different species are under-recorded at the 'average' UKBMS site. We stress these are preliminary estimates based on relatively few sites and, in some cases, on small sample sizes. Notwithstanding these caveats, the numbers suggest that several species of UK conservation concern are being systematically under-recorded: the Dingy skipper, Grayling and Silver-studded blue (see also Dennis & Sparks 2006) are all priorities on the UK Biodiversity Action Plan and are among the least detectable of the 17 species studied here (Fig. 2). Among species, both colour and size have limited power in explaining detectability, although the relative
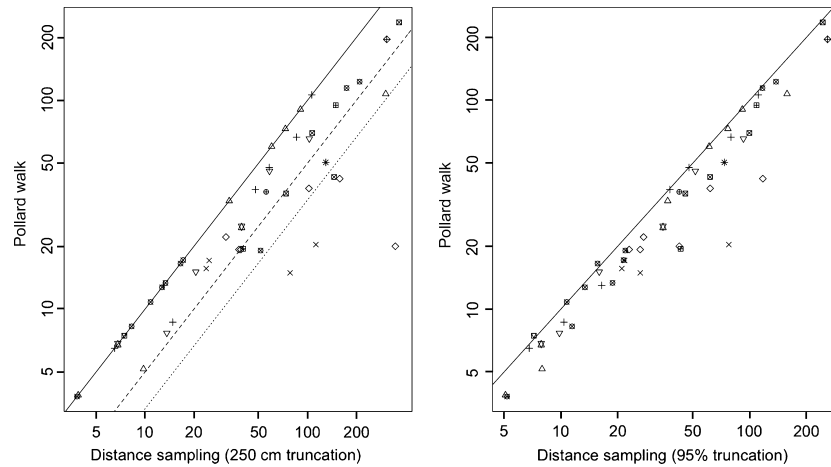
Fig. 3. Comparison of population density (individuals per hectare) estimated by the Pollard Walk and distance sampling, using both a 250-cm truncation (left panel) and the 95% truncation (right panel). Each symbol represents a different study site. The solid line indicates the 1:1 relation-
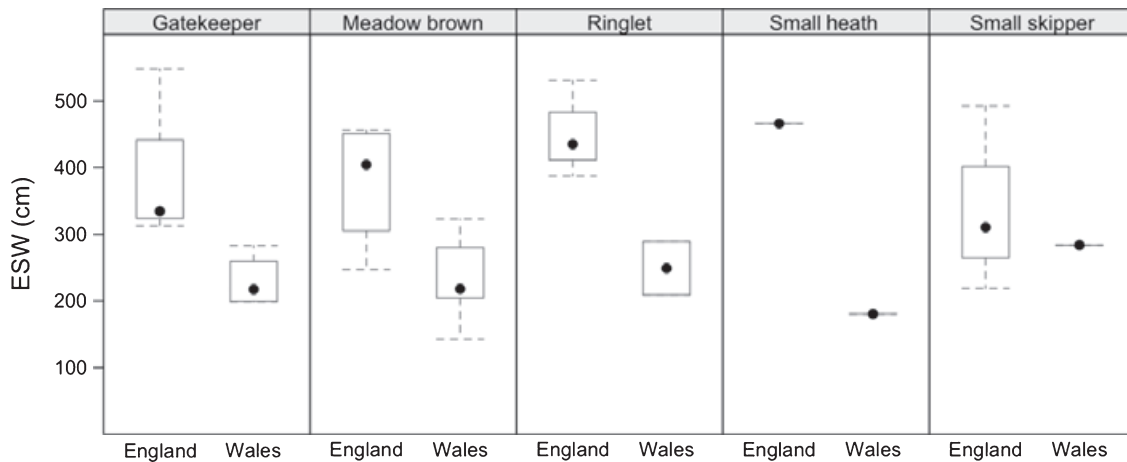

Fig. 4. Box-and-whiskers plot showing variation among sites in effective strip widths (in cm) for species observed at sites in both England and Wales. Data derived from data in 10 equally spaced bins after truncating at the 95% of observations for each site–species combination.

position of most species on this gradient of detectability is not surprising: the Dingy Skipper and Grayling are both well-camouflaged and known to be difficult to spot, whilst the three pierid species are all highly conspicuous.

Site effects are at least as important as species identity in determining detectability. Within the Pollard box, variance in detectability is much greater among sites than among species, which suggests that any correction factor applied to UKBMS data should be biotope-specific as well as species-specific. Our variance components model predicts the correction factor for the 'average' species to be in the range 1·1–2·5 for 95% of sites; comparable prediction intervals for species at the 'average' site are 1·3 – 1·9. This suggests that UKBMS data might not be especially reliable for comparing butterfly abundance between sites in individual years. However, the 2·5-fold variation in detectability remains small compared with the 100-fold variation in estimated abundance that is typical of species on the UKBMS (Thomas, Simcox, & Hovestadt 2011).

The importance of the site effect is evident in the left-hand panel of Fig. 3: most of the severely underestimated population densities are found at just a few sites (principally the Dulas Valley sites). The lack of significant relationship between vegetation height and detectability suggests that sites differ in ways that are not captured by our simple index, especially because butterfly behaviour varies among biotopes (Dennis 2004) in ways that have unpredictable consequences for detectability. The strongest pattern in detectability is that detection distances in at Welsh sites were substantially shorter than in English ones. This could be explained by the coastal location and therefore higher wind speeds in Wales (wind makes it difficult to identify butterflies, especially in flight). However, the studies were conducted on different butterfly species at different times and by different observers. We can reject the effect of species composition, as the regional difference is pronounced among four of the five species shared between English and Welsh study sites (Fig. 4). The survey year is potentially confounding,

because the English data were collected during an extremely hot summer (2006), whilst the Welsh study was conducted during a relatively poor year for butterflies (1998). The UKBMS minimum weather conditions (Pollard & Yates 1993) were observed during both studies presented here, but it is likely that variation in weather above these minima exerts a strong influence on butterfly behaviour that have knock-on effects for detectability (Dennis & Sparks 2006; Wikstrom, Milberg, & Bergman 2009). The final complication is that two different observers collected the data. Both observers received suitable training and it seems unlikely that differences in their ability to identify butterflies and estimate distances can account for the much larger ESWs at sites in England. Variation among observers presents greater problems for extrapolating our results to the wider question of detectability. Both our observers were relatively naïve: more experienced recorders might have an established search image of species of particular conservation concern, even if they are difficult to see. Such experience almost certainly increases the detectability of species with distinctive flight patterns (e.g. dingy skipper) but also presents an extra source of variation. Variation among recorders therefore deserves further consideration (Kery & Plattner 2007; Nowicki *et al.* 2008), possibly by observing a range of recorders surveying the same sites. The importance of intraspecific variation in detectability means that untangling these multiple causal factors is a priority for future research in this area.

We found tight correlations between densities estimated using the Pollard Walk and distance sampling (Fig. 3). This is because variation in detectability, whilst substantial, is small compared to the huge variation in population density across sites and species (c.f. Thomas, Simcox, & Hovestadt 2011). However, for some populations, the Pollard Walk gives a substantial underestimate. Thus, it would be unwise to treat Pollard Walk data as absolute estimates of abundance without considering the factors correlated with detectability. Most existing applications of butterfly monitoring scheme data are based on trends over time within populations (Roy & Sparks 2000; Roy *et al.* 2001; Warren *et al.* 2001; Brereton *et al.* 2007): the key question here is whether the variation in detectability within populations is of comparable magnitude to real changes in population size. The widely reported trends in butterfly abundance (e.g. Fox *et al.* 2006) might be compromised if biotopes themselves had changed in a consistent way across the country over the period of monitoring, thus leading to a systematic trend in detectability. National-scale trends are probably quite robust, given that species declining on the UKBMS tend also to have shrinking distributions (Warren *et al.* 2001; Thomas 2005; Fox *et al.* 2006), but individual site-level trends might be less precise. Long-term vegetation change might conceivably increase detectability (making it harder to detect declines in abundance) or decrease it (making it appear that stable populations are in fact declining). We suspect that the observed inter-site variation is far greater than the likely range for any one site, even under the combined effects of ecological succession, management, weather and climate change. However, this unanswered question could be addressed by a combination of monitoring detectability at reference sites and controlled experiments that manipulate biotope structure in realistic ways. Such focussed research should use MCDS (Thomas *et al.* 2010) rather than the stratified approach employed in this study.

Although we have demonstrated the value and potential of distance sampling in butterfly monitoring, there are both practical and theoretical considerations that make distance sampling unviable as an alternative to the Pollard Walk for large-scale multi-species monitoring. The practical issue is the potentially large number of butterflies occurring in peak season, when it is commonplace to record a butterfly every second. The effort of keeping separate counts for each species is so intense that it would be impossible to record distance estimates for each observation, even in the wider countryside. More fundamentally, most animals tend to be observed in flight, which violates one of the key assumptions of distance sampling (but see Buckland *et al.* 2001 p198). In addition, UKBMS routes do not sample habitat randomly, either at small spatial scales (many routes follow linear features or public rights of way) or large (sites tend to be selected because they contain abundant populations), leading to biased estimates of population density (either from distance sampling or the Pollard Walk). Our detection function models were hampered by the fact that several transects followed paths, which provide warm microclimates that attract aggregations of basking butterflies, thus violating another key assumption of distance sampling. Unfortunately, it would be impractical and undesirable to relocate traditional UKBMS transects to be more representative without breaking the continuity of > 3 decades continuous monitoring that is the major strength of the scheme. The UK monitoring has recently been extended through a complementary scheme, the Wider Countryside Butterfly Survey, that samples a stratified random selection of survey locations (1 km grid squares) across the UK (Roy, Rothery, & Brereton 2007; Brereton *et al.* 2011). Although the wider countryside scheme addresses the bias towards sampling high abundance sites, it still involves routes that follow linear features or public rights of way. In spite of these reservations, we suggest that distance sampling, particularly MCDS, has two important roles in butterfly monitoring. One is to conduct intensive studies on a species-by-species basis to refine our estimates of detectability and quantify the importance of variation because of biotope, management conditions, weather conditions, observer, butterfly behaviour (perched vs. flying) and sex (across the three species of Blue butterflies, 90% of observations were on males). The second is to conduct targeted surveys and monitoring in relatively open biotopes, where trained observers can collect data outside the 250 cm of the Pollard Walk box. This approach would be especially suitable for species of high conservation concern (e.g. Large Blue and High Brown Fritillary), where absolute abundance estimates may be important for conservation and research.

The work described here is not the final word on detectability of butterflies on transects but provides an important step in testing the robustness of Pollard Walk data (see also Haddad *et al.* 2008; Nowicki *et al.* 2008). Monitoring schemes like the UKBMS are increasingly being used to address questions

about global change (de Heer, Kapos, & ten Brink 2005). Validation of these data, using well-established ecological methodology, is therefore essential for delivering policy objectives for biodiversity, both nationally (Sutherland *et al.* 2006) and internationally (Dobson 2005). With this in mind, we hope that our work will provoke new enquiry into methodological questions about biodiversity change and contribute to the development of more rigorous standards in applied ecology and conservation (Sutherland *et al.* 2004; Stewart 2010).

Acknowledgements

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Supporting Information

Additional Supporting Information may be found in the online version of this article.

Fig. S1. Effective strip widths as a function of the number of bins.

Table S1. Details of the distance models fit to each of 50 site-species combinations.

Table S2. Covariate detection model comparisons for data truncated at 250 cm (left side) and data truncated at the 95th percentile of distances (right side).

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