

Morphological variation associated with dispersal capacity in a tree-killing bark beetle *Dendroctonus ponderosae* Hopkins

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ABSTRACT

1. Intrinsic factors influencing dispersal of insect pests during outbreaks are poorly understood, yet these factors need to be quantified to parameterize dispersal in models that predict population spread.

Our study related wing and body morphology of female mountain pine beetles (*Dendroctonus ponderosae*) to flight distance, as measured by flight mill bioassays.

2. Beetles that flew long distances (> 11 km) had greater body weight and larger wings than beetles that flew short distances (< 1 km). These heavier female beetles should also be more capable pioneers, as other studies have shown that body weight is positively correlated with lipid content.

3. Wing and body morphology of females are significant predictors of flight distance; heavy beetles with large wings generally flew further than smaller beetles, although this relationship is heteroscedastic. Dispersion of flight distance values increases with wing loading (weight/wing area) due to a cohort of 'lazy' individuals that fly short distances (< 1 km) regardless of flight phenotype.

4. Observed morphology explained less than 20% of the variation in flight capacity, indicating substantial contribution from other intrinsic factors that remain to be investigated. This study may have implications for dispersal modeling, providing estimates of flight capacity using morphological measurements.

Keywords: bark beetle, *Dendroctonus ponderosae*, morphological variation, dispersal kernel, flight, wing morphology, dispersal modeling, body weight, heteroscedasticity, pest

1 INTRODUCTION

2 Dispersal leading to range expansion can exert selective pressures on the physical traits associated with
3 dispersal phenotypes (Olivieri and Gouyon, 1997; Hassall 2015). In many insects, traits like wing
4 morphology, flight muscle and body condition can influence flight capacity and dispersal potential
5 (Azevedo *et al.*, 1998; Hassall, 2015). Measurements of such flight-related traits can be used to infer an
6 individual's ability to disperse, and may also be correlated with successful establishment after dispersal
7 (Manning and Reid, 2013; Evenden *et al.*, 2014). Models that predict dispersal capacity by pest species
8 have become important tools for environmental conservation and resource management through the
9 refinement of damage prevention, especially in cases of range expansion and invasion of novel habitats
10 (Safranyik and Carroll, 2006).

11 The mountain pine beetle (MPB) *Dendroctonus ponderosae* Hopkins (Coleoptera: Curculionidae) is a
12 tree-killing bark beetle that attacks and kills pine trees across western North America. MPB has recently
13 expanded its range eastward (Safranyik *et al.*, 2010). This spread is threatening to continue; historically,
14 MPB distribution has been limited by climate rather than host availability (Safranyik, 1978), but climate
15 change has made previously unsuitable regions available to this species (Carroll *et al.*, 2004). Milder
16 climate and protection of large overmature trees from natural fires have created conditions in which the
17 beetles can thrive and reproduce to reach epidemic population levels (Taylor and Carroll, 2004;
18 Safranyik and Carroll, 2006). This has led to the destruction of over 723 million m³ of merchantable pine
19 in British Columbia and Alberta during the last outbreak (NRCan, 2017), with subsequent ecological and
20 economic impacts (Kurz *et al.*, 2008; Corbett *et al.*, 2015). These losses can be reduced through
21 effective allocation of management resources based on early risk detection (Safranyik and Carroll,
22 2006). Resource allocation is increasingly important in epidemic scenarios in which total control of MPB
23 is not realistic. In such scenarios, equipment and man-power must be allocated in accessible areas that
24 will maximize the impact on MPB spread while minimizing cost. Predictive modelling of MPB spread is

25 one means of risk detection, and such models need to be parameterized with variables such as regional
26 climate and dispersal capacity (Atkins, 1961; Robertson *et al.*, 2007; Aukema *et al.*, 2008; Goodsman *et*
27 *al.*, 2016). Climatic data are often available, but the ability to identify MPB dispersal phenotypes is
28 lacking.

29 The variable dispersal behaviour of MPB poses a challenge for predicting its range expansion. Beetles
30 can utilize updrafts to disperse above the forest canopy (Jackson *et al.*, 2008; de la Giroday *et al.*, 2012),
31 but, research has shown that a majority of dispersal can take place below the canopy (Safranyik *et al.*,
32 1992). In below-canopy dispersal, many beetles ignore susceptible host trees immediately upon
33 emergence in order to colonize new locations (Safranyik *et al.*, 1989; Robertson *et al.*, 2007). Below-
34 canopy dispersal is self-propelled and can be parameterized in vitro with computer-linked flight mill
35 bioassays. Many models predict MPB dispersal at both landscape and stand levels based on spatio-
36 temporal patterns (Robertson *et al.*, 2007; Aukema *et al.*, 2008; Chen and Walton, 2011), climate
37 (Aukema *et al.*, 2008), host vigor (Lewis *et al.*, 2009) and other environmental conditions (Safranyik *et*
38 *al.*, 1989; Robertson *et al.*, 2007; Ainslie and Jackson, 2011). Inclusion of flight mill data can improve
39 model accuracy by providing detailed data on flight behavior and patterns such as stopping frequency
40 and length of uninterrupted flight times (Goodsman *et al.*, 2016). Finding the relationships of MPB wing
41 and body morphology to measured flight capacity should improve our ability to predict bark beetle
42 dispersal in nature.

43 In insects, wing morphology (*e.g.*, shape and size) and flight muscle mass are often correlated with
44 dispersal capability. Factors such as variance in aspect ratio (ratio of wing length to width) have been
45 linked to greater dispersal in populations at higher latitudes (Azevedo *et al.*, 1998; Hassall, 2015). For
46 example, small wing aspect ratio and large relative wing size can increase the flight capacity of
47 *Drosophila* in cool climates (Azevedo *et al.*, 1998; Frazier *et al.*, 2008). In several insect species, such as
48 milkweed bugs, neotropical butterflies and sand crickets, there are positive relationships between wing

49 span and flight frequency, duration and speed (Palmer and Dingle, 1989; Dudley 1990; Fairbairn and
50 Roff, 1990). Along with wing size, thoracic muscle mass can influence flight capability in butterflies
51 (Dempster *et al.*, 1976; Chai and Srygley, 1990). In bark beetles, body mass correlates with flight
52 distance and duration (Williams and Robertson, 2008; Evenden *et al.*, 2014). In general, body mass has
53 been a primary factor in body condition quantification (Elkin and Reid 2004; Elkin and Reid, 2005). Body
54 condition is considered to represent the intrinsic resources available for growth, survival and
55 reproduction, and is consequently important for successful colonization (Boggs, 1992).

56 Dispersal is necessary in MPB as adults must find a new host tree in order to reproduce (Safranyik and
57 Carroll, 2006), and females are pioneers in host colonization. Females initiate the attack once they
58 arrive on a host and release aggregation pheromones to attract conspecifics (Blomquist *et al.*, 2010),
59 and bark beetle body condition determines the quality and quantity of their offspring (Kautz *et al.*,
60 2016). Correlations among lipid content, attack success and reproduction in bark beetles have been
61 well documented (Elkin and Reid, 2005; Reid and Purcell, 2011; Manning and Reid, 2013; Kautz *et al.*,
62 2016); larger, heavier beetles have greater lipid content (Evenden *et al.*, 2014), resulting in increased
63 fecundity (Elkin and Reid, 2005), and greater ability to resist host defences (Reid and Purcell, 2011).
64 Larger females therefore are potentially better pioneers through their capacity to facilitate range
65 expansion.

66 As the condition of females is influential in colonizing novel regions, this study aims to identify and
67 quantify the morphological factors associated with female MPB dispersal capacity. These morphological
68 features of females could provide a way to characterize the flight potential of individuals and dispersal
69 capacity of populations (Hill *et al.*, 1999b; Hassall *et al.*, 2009). Given that flight performance is often
70 associated with phenotypic traits such as wing size and body condition, larger individuals with larger
71 wings should have greater dispersal capacity.

72

73 **METHODS**

74 **Beetle data collection**

75 All MPB used for this study were collected near Grande Prairie, Alberta, Canada (55.17 N, 118.80 W).

76 The sample group was reared from bolts collected in October 2015. Four lodgepole pine trees infested
77 by MPB were selected (3 from site 1 at 54.57 N, 119.42 W; and 1 from site 2 at 54.19 N, 118.68 W), and
78 two 50-cm bolts were cut from each tree at 1m above ground level. All bolts were sealed with paraffin
79 wax on the cut ends upon felling. These bolts, containing larval MPB, were stored at 4°C until use.

80 Bolts were removed from 4°C and placed in separate 136 L opaque plastic emergence chambers at
81 24±1°C in April 2016. Emerged beetles were collected daily. A total of 173 females were collected and
82 flown; however, 49 beetles that flew less than 10 m, 44 of which did not fly at all, were removed from
83 this study in order to exclude individuals that had irregular flight behaviours that were unlikely to be
84 related to observed morphological features. Overall, the experimental group consisted of 124 female
85 beetles, including 53 collected from site 1 bolts and 71 from site 2 bolts. Individual beetles were
86 separated by sex based on the incidence of beetle stridulation, which is considered to be ~98% accurate
87 for identifying males (Rosenberger *et al.*, 2016). This was performed at 24±1°C. The beetles were then
88 placed in 2 mL centrifuge tubes with a small piece of paper, and stored at 4°C to reduce use of metabolic
89 resources until the flight assay (Evenden *et al.*, 2014), which was performed 3 to 5 days post emergence.

90 Flight distance, duration and frequency data was collected for female beetles using the methods for
91 computer-linked flight mill bioassays of Evenden *et al.* (2014). Flight distance was calculated as the total
92 distance flown over a flight period, flight duration was calculated as total time spent in-flight, and flight
93 frequency was calculated as the number of times flight was initiated after a minimum 5 second period of
94 non-flight. Females were weighed prior to flight (Mettler Toledo, XS105 accurate to 0.01mg), and body

95 dimensions (total length and pronotum width) were measured using digital calipers (Pro.Point 1-
96 150mm, accurate to 0.01mm; remeasuring of 10 randomly selected specimens verified 99.01%
97 precision). Beetles were then attached to a 2 cm long tether of 0.32 mm-diameter aluminum wire using
98 LePage® Heavy Duty Contact Cement. Tethers were carefully attached to the pronotum to ensure no
99 interference of elytra and wing movement. Length of the flight bioassay was 22 hours. This duration
100 was chosen to allow for processing of samples and switch over of specimens between flight cohorts.
101 During the flight assay the flight mill chamber had a photoperiod of 16:8 (L:D) h and a temperature of
102 22.5 °C. Flight assays began 2 hours after initiation of the light phase of the photoperiod, which allowed
103 time to process the previous flight group, and to prepare the next group of beetles for the 22-hour flight
104 assay. Female beetles were flown in groups ranging from 8 to 20.

105 After flights were complete, beetles were weighed again and stored in 85% ethanol at -20°C. For
106 morphological data collection, wings were removed and suspended in a 50% ethanol solution. Wings
107 were unfolded using paintbrushes and positioned, then removed from the solution by sliding a small
108 strip of card stock below the wing and gently lifting it, maintaining the unfolded wing shape. The wings
109 were fixed to the card stock strip by applying a thin layer of diluted glue with a fine paintbrush. Dried
110 wings were scanned and measured using ImageJ version 1.51j8 (Schneider *et al.*, 2012).

111 Sixty-two MPB individuals were available for muscle dissection. The metathoracis medianus, lateralis
112 posterior and lateralis medius flight muscles were identified for dissection based on Reid (1958) and
113 Chapman (1998). After removal of the wings and elytra, the metathorax was opened by removing the
114 dorsal tergites. Once the flight muscles were exposed, fibre length and lateral width of the metathoracis
115 medianus were measured using an ocular micrometer in a dissection microscope under 50x
116 magnification. The anterior edge of a medial segment of the metathoracis medianis was then detached
117 from the metathoracic prephragma and its transverse thickness was measured. Subsequently, the
118 transverse thickness and lateral width of the lateralis medius and the lateralis posterior were measured.

119 The lateralis medius was then carefully separated from the metacoxa, extracted from the metathorax,
120 placed flat and its length measured. Measurements of the volume of individual muscles were then
121 summed into 2 groups: dorsal longitudinal flight muscle volume, which included the metathoracis
122 medianus muscle group, and dorsal ventral flight muscle volume, which included measurements of the
123 lateralis medius and the lateralis posterior. Individual muscle measurements were each included in
124 preliminary models, but were not found to provide additional explanatory power individually. To
125 simplify variables and interactions in the presented models, these measurements were consolidated by
126 calculating the volume of an elliptical cylinder, representing the flight muscle volume, and this value was
127 used for further analyses.

128 **Statistical Analysis**

129 All statistical analyses were conducted in R version 3.3.2 (R Core Team 2017). Model R^2 values were
130 obtained using the R package piecewiseSEM (Lefcheck, 2015) which finds R^2 values for non-fixed effect
131 linear models, based on Nakagawa and Schielzeth (2013) and Johnson (2014). Flight models were
132 created and analysed using R packages nlme (Pinheiro *et al.*, 2017) and lme4 (Bates *et al.*, 2015).

133 ANOVA hypothesis testing was used to determine whether differences exist between the strongest and
134 weakest fliers observed in this study. Samples were separated by quartiles based on flight distance in
135 order to apply the ANOVA tests. The upper quartile beetles, with flight distances > 11.13 km (n=31),
136 represented strong fliers. They were compared to the lower quartile beetles that represented weak
137 fliers, with flight distances < 1.05 km, (n=31; this number excludes any beetles that flew less than 10 m).
138 Muscle volume was also compared using ANOVA hypothesis tests on the top and bottom quartiles, but
139 with smaller sample sizes since fewer beetles with dissected flight muscle were available. The top
140 quartile group contained 16 beetles, with flight distances > 8.0 km, and the bottom quartile group
141 contained 16 beetles, with flight distances < 3.5 km. Morphological features tested included body

142 weight, body size (calculated as ellipsoid volume [$\frac{4}{3} * \pi * body\ length * pronotum\ width^2$]), preflight
143 weight, postflight weight, density ($\frac{preflight\ weight}{body\ size}$), absolute weight lost, proportional weight lost, wing
144 size (measured as two-dimensional surface area), and wing shape (calculated as aspect ratio:
145 $\frac{wing\ length}{wing\ width}$).

146 A linear mixed effects model was produced using distance flown as the dependent variable. Each
147 included emergence bolt and flight mill as random variables. Each full model included the independent
148 variables: pre-flight weight, body size, wing size, age (days since emergence) and wing shape, as well as
149 all two-way interactions. Step-wise model simplification was achieved by removal of the least significant
150 variables (highest p-values) until only significant ($p < 0.05$) variables remained, and reduced models were
151 compared to previous models and the full model using AIC and ANOVA hypothesis testing until the most
152 parsimonious model was found. Following model refinement methods by Crawley (2013), interactions
153 between variables were preferred for removal over basic variables in instances where both variables
154 and interactions had a p-value greater than 0.05. All models were tested for significance using F-tests,
155 and the included independent variables were tested with ANOVA. The residuals of each regression
156 model were observed for random dispersal, and a variance inflation factor test was performed on each
157 model before and after refinement using the R package *fmsb* (Nakazawa, 2017)

158 We tested the significant regression variables of distance flown for heteroscedasticity using Breusch-
159 Pagan tests. When heteroscedasticity was present, quantile regressions were used to regress the data
160 at separate quantiles of the independent variable, using the R package *quantreg* (Koenker, 2016).

161 Because variables can only be tested independently in quantile regression, a third variable, wing loading
162 ($\frac{preflight\ weight}{wing\ area}$), was introduced to combine the significant independent variables into a single term.

163 Tests were conducted at the quantiles of 0.12 to 0.96 in increments of 0.12 in order to maximize

164 resolution while retaining an adequate number of samples per quantile. When heteroscedasticity was
165 observed, ANOVA hypothesis testing was used to compare slopes to an average quantile slope.

166 RESULTS

167 Morphology Associations

168 Variability in the following data is reported as standard error. Female MPB flew an average distance of
169 7.4 ± 0.58 km over the 22-hour flight period, with a maximum flight distance of 28.81 km. On average,
170 beetles flew for 3.70 ± 0.29 hours and initiated flight 176 ± 27.52 times; 81.8% of the total flight
171 distance occurred during the light stage of the photoperiod. Strong fliers flew longer, and more
172 frequently, than weak fliers ($F_{(1,60)}=402.230$, $P<0.0001$; and $F_{(1,60)}=7.433$, $P=0.0084$, respectively). Strong
173 fliers flew an average of 7.60 ± 0.36 hours with an average frequency of 144.8 ± 29.65 flights, while weak
174 fliers flew an average of 0.170 ± 0.03 hours with 55.5 ± 12.67 flights. Forty-four beetles did not fly at all,
175 while 5 flew less than 10 m. Dispersal distance data was not normally distributed (Figure 1).

Insert Figure 1

176 Long distance and short distance fliers were morphologically different (Figure 2). Beetles that flew
177 longer distances had greater body weight, both preflight ($F_{(1,60)}=9.021$, $P=0.0039$) and postflight
178 ($F_{(1,60)}=6.637$, $P=0.0125$), as well as greater body density ($F_{(1,60)}=15.741$, $P=0.0002$). Long distance fliers
179 also lost more absolute weight ($F_{(1,60)}=10.990$, $P=0.0016$) and proportional weight ($F_{(1,60)}=5.467$,
180 $P=0.0227$) during the flight bioassay. Long distance fliers tended to have larger wings, but this only
181 bordered significance ($F_{(1,60)}=3.740$, $P=0.0578$; Figure 2). There was no significant difference between
182 long and short distance fliers in wing shape ($F_{(1,60)}=1.453$, $P=0.2328$), body size ($F_{(1,60)}=1.332$, $P=0.2530$),
183 or flight muscle volume ($F_{(1,30)}=0.1008$, $P=0.7530$). Plots containing all 124 samples showing the
184 relationship between total distance flown and preflight weight, postflight weight, wing area, body
185 density, proportional weight lost and absolute weight lost can be seen in Supplementary Figure S1. The
186 age of beetles in this study (3-5 days) had no relationship with flight distance ($F_{(1,60)}=0.440$, $P=0.5095$).

Insert Figure 2

187 **Regression Models**

188 After model refinement, flight distance of MPB was significantly related to both wing size ($F_{(1,121)}=8.512$,
189 $P=0.0042$) and preflight weight ($F_{(1,121)}=12.466$, $P=0.0006$). Heavier beetles with larger wings flew farther
190 than lighter beetles with smaller wings. Wing data was essential in predicting flight performance;
191 exclusion of wing size data resulted in a significant loss of explanatory power in the flight distance model
192 ($F_{(121,122)}=8.512$, $P=0.0042$), as well as an increase in AIC (AIC 2516.0 to 2522.4). Ultimately, these linear
193 mixed effects models explained only a limited amount of the flight performance in female MPB. The
194 models predicted 19% of variation in flight distance ($R^2=0.1910$).

195 A variance inflation factor (VIF) of 10 or more indicates multicollinearity; VIF tests showed that there
196 was no multicollinearity before or after model refinement (VIF=1.32 and VIF=1.17, respectively).

197 **Heteroscedasticity and Quantile Regression**

198 Breusch-Pagan tests showed potential heteroscedasticity in the relationship of distance flown to
199 preflight weight ($BP_{(1)}=10.447$, $P=0.0012$) and wing area ($BP_{(1)}=4.028$, $P=0.0447$), which were more
200 prevalent when both variables were tested simultaneously as wing loading ($BP_{(1)}=16.811$, $P<0.0001$).

Insert Figure 3

201 Preflight weight and wing area were first tested with quantile regression separately (Figure 3). Distance
202 flown had a significantly different response to preflight weight at the 96% quantile ($F_{(1,247)}=5.399$,
203 $P=0.0210$), but showed no significant difference at the 12% quantile ($F_{(1,247)}=3.369$, $P=0.0676$). Preflight
204 weight had a significantly stronger effect on flight distance when it was greater, with an increase of 1.63
205 km in flight distance per 1 mg increase in beetle weight, while it otherwise had an average increase of
206 0.67 km per 1 mg increase in preflight weight.

207 When both variables were combined into wing loading, the tests for heteroscedasticity revealed a
208 general trend in which the dispersion of distance flown values increased with wing loading.

Insert Figure 4

209 Distance flown had a significantly different response from the average at both the 12% quantile
210 ($F_{(1,247)}=12.132$, $P=0.0006$; Figure 4 and the 96% quantile ($F_{(1,247)}=4.495$, $P=0.0350$; Figure 4). At the 96%
211 quantile, the predicted distance flown increases by 38.5 km per 1 mg/mm² increase wing loading, while
212 at the 12% quantile this relationship was reduced to an increase of 1.6 km per 1 mg/mm². Otherwise,
213 the average increase was 19.3 km per 1 mg/mm² wing loading.

214 **DISCUSSION**

215 **Morphological Associations with Dispersal Capacity**

216 We found that heavier beetles with larger wings flew farther than lighter beetles with smaller wings.
217 After simplification, linear mixed effects models showed that the primary variables explaining distance
218 flown were wing size and preflight body weight. Increased dispersal capacity by heavier individuals has
219 also been found in other studies (Hill *et al.*, 1999a; Williams and Robertson, 2008; Evenden *et al.*, 2014).
220 Although body weight was a significant predictor of dispersal in this study, body size was not related to
221 flight distance. Reid and Purcell (2011) found no relationship between body size and fat content, but
222 demonstrated that relative body weight is correlated with fat content. Other research has also shown
223 the relationship between weight and lipid content (Williams and Robertson, 2008; Evenden *et al.*, 2014),
224 implying that denser beetles have a greater store of flight fuel per unit of body size that allows them to
225 fly further. This is also supported by our findings of increased density in long distance fliers.

226 An increase in stored lipids can also influence dispersal behaviour in bark beetles through behavioural
227 responses to olfactory stimuli. Some newly emerged beetles will ignore both visual and chemical cues
228 from susceptible hosts (Shepherd, 1966; Safranyik *et al.*, 1989; Eidson *et al.*, 2017). In the Douglas-fir
229 beetle, *Dendroctonus pseudotsugae*, beetles with higher fat content are less responsive to aggregation
230 pheromones and spend more time selecting a host to colonize (Atkins, 1966; Bennett and Borden,
231 1971), thereby increasing dispersal; depletion of fat reserves through flight initiates a behavioural

232 response in which the beetles are more responsive to olfactory cues. Fat content likely has a similar
233 effect on dispersal in MPB, as well as impacting colonization success (Seybold *et al.*, 2006; Reid and
234 Purcell, 2011) and reproduction (Manning and Reid, 2013) after dispersal.

235 We have shown that long range dispersing beetles not only have higher preflight weight, but postflight
236 weight as well, which can influence the success of pioneers. Successful colonization requires large
237 amounts of stored fat to detoxify monoterpenes produced by the host tree (Reid and Purcell, 2011), and
238 to produce offspring (Elkin and Reid 2005; Manning and Reid, 2013). Greater postflight weight provides
239 a substantial advantage to MPB, since the heaviest individuals are most likely to disperse long distances
240 and also have the greatest chance of succeeding in host colonization.

241 While weight is important in dispersal capacity of MPB (Evenden *et al.*, 2014), the inclusion of wing
242 morphology significantly increased the predictive power of the regression models tested in this study.

243 Wing size is often important for flight performance in insects (Betts and Wootton, 1988; Taylor and
244 Merriam, 1995; Hill, 1999a), and larger wings increase lift and carrying capacity (Wootton, 1992),
245 allowing longer distance travel with a greater load of resources for dispersal and colonization, such as
246 lipids. Large wings are also important for flight capacity in cold climates; cold temperatures can reduce
247 flight muscle function and wing beat frequency, but larger wings allow for increased lift generation with
248 lower input power (Azevedo *et al.*, 1998; Frazier *et al.*, 2008). Individuals with such an advantage could
249 have contributed to the recent range expansion of MPB north into the Northwest Territories (NRCan,
250 2017), and eastward across the Rocky Mountains, although this hypothesis remains to be tested.

251 The ability to predict range expansion can be challenged by difficulties in acquiring data for predicting
252 dispersal phenotype; however, morphological data related to dispersal capacity can be obtained from
253 specimens that are already available from population monitoring. Currently, 4-inch diameter disks are
254 collected from infested trees in many mass attack regions, and larvae are observed under the bark to

255 determine the reproductive success of a population (Alberta Agriculture and Forestry, 2016).
256 Quantification of the relationship between larval instars and adult size is needed, but, with some
257 modification of sampling procedure, collected data, such as larval weight, could be incorporated into
258 this data collection and subsequently used in dispersal models that distinguish between regional
259 populations.

260 Although not investigated in this study, wing size may also affect above-canopy passive dispersal. Range
261 expansion in MPB can also occur through long distance dispersal events in which beetles are caught in
262 updrafts and carried up to 300 km (Cerezke, 1989; Jackson *et al.*, 2008). Certain behaviours exhibited by
263 MPB hint at adaptations to passive dispersal: 44 of the beetles that were tethered to flight mills did not
264 actually fly, but many of these displayed a “drifting” behaviour in which they would open the elytra and
265 spread their wings fully, without flapping. Similar behaviors have been documented by Atkins (1959)
266 when testing flight preparation and response in the Douglas-fir beetle, *D. pseudotsugae*. They found
267 that some beetles fully extended their wings, but did not vibrate them, when tossed in the air. We
268 suggest that such behaviour, when caught in an updraft, could maximize the exposed wing area while
269 minimizing energy consumption. Associations between stationary wing size and passive dispersal
270 distance occur in winged seeds of plants, (Augspurger, 1986) and this relationship could be similar in
271 insects dispersed by the wind.

272 **Heteroscedasticity in Flight Distance and Future Directions**

273 Many factors can affect flight performance, and although the morphological variation observed in our
274 study follows a normal distribution, the distribution of flight capacities follow a well-known non-normal
275 pattern of dispersal, the dispersal kernel (Figure 1; Bateman, 1950; Chapman *et al.*, 2006). We found
276 that many individuals fly only short distances on the flight mill while very few fly long distances. There
277 are population-wide advantages to this distribution of dispersal behaviours, as it gives the greatest
278 survivorship in the largest portion of the population while still allowing some risk for colonization.

279 Beetles that fly short distances use less lipid (Williams and Robertson, 2008; Evenden *et al.*, 2014) and
280 tend to be more successful in colonizing host trees than similarly sized beetles that burn more of their
281 lipid reserves to disperse (Reid and Purcell, 2011; Manning and Reid, 2013; Kautz *et al.*, 2016). In our
282 study, we found that body weight and wing size were associated with flight distance and, consequently,
283 a population of larger individuals with larger wings is likely to have a greater proportion of long distance
284 dispersers; however, the strong positive skew in the dispersal kernel of MPB causes inconsistencies with
285 predictions based on linear relationships with morphology.

286 Even with both wing and body morphology included, the power of these models to predict flight
287 distance remains relatively low. Part of the weakness in the relationship of morphology with flight
288 distance is due to heteroscedasticity in the coefficients of the variables. Quantile regression shows that
289 preflight weight and wing area, when considered separately, do not show a clear relationship with flight
290 distance (Figure 3); however, when combined as wing loading, the heteroscedastic nature of the
291 relationship becomes evident (Figure 4). When wing loading is low, morphology has little power to
292 predict dispersal, but the relationship at high wing loading values has a significantly higher slope,
293 indicating that small increases can have a significant positive impact on dispersal capacity (Figure 4).

294 This inconsistent relationship is caused by an increase in the dispersion of flight distance at different
295 quantiles of wing loading, and is likely due to the large cohort of 'lazy' individuals that fly only short
296 distances regardless of high wing loading (Figure 4). These 'lazy' beetles increase the disparity between
297 individuals with low and high wing loading within each quantile, which subsequently increases the
298 quantile slope coefficient. This creates difficulties in predicting the relationship between morphology
299 and flight capacity, and indicates that there is more than morphology associated with dispersal capacity.

300 The natural tendency to disperse is an important contributor to flight distance in other insects (Steyn *et*
301 *al.*, 2016), and further research should focus on identifying factors that contribute to variation in
302 dispersal behaviour, such as the 'lazy' dispersal phenotype seen in our study.

303 Other factors involved in dispersal capacity in the natural environment that could not be tested in this
304 lab study include abiotic conditions such as temperature (Atkins, 1959; Aukema *et al.*, 2008), wind
305 (Safranyik *et al.*, 1989; Ainslie and Jackson, 2011), and humidity (Safranyik and Carroll, 2006). Biotic
306 factors that can influence dispersal include host presence, density and distance from the beetle (Atkins,
307 1966; Robertson *et al.*, 2007), wing flexibility (Mountcastle and Combes, 2013), joint resilin content
308 (Haas *et al.*, 2000), as well as parasitic infections (Everleigh *et al.*, 2007). Some of the variation in the
309 relationship between beetle morphology and flight performance may be mediated by behaviour. Atkins
310 (1966) has shown that different female dispersal behavioural types exist within Douglas-fir beetle
311 populations, but the behaviours exhibited by these types did not appear to have a strong correlation
312 with body weight. Our results similarly reveal that some 'lazy' individuals do not fly to their capacity as
313 indicated by wing size and body weight. Although morphological characteristics demonstrably have an
314 impact on flight capacity in MPB, other factors that govern dispersal-related behaviour, such as gene
315 expression, should also be investigated.

316 **Conclusions**

317 Body weight and wing size contribute to dispersal capacity in mountain pine beetle, and such data can
318 potentially be useful in improving dispersal model quality (Goodsman *et al.*, 2016). Although many
319 other factors must also be considered in the parameterization of dispersal models, we have shown that
320 large female beetles with large wings generally have increased dispersal capacity, flying farther than
321 smaller beetles with smaller wings; autonomous populations in dispersal models can be assigned flight
322 capacities based on morphology observed *in situ* through use of samples already being collected for
323 monitoring purposes.

324 As large females are better colonizers (Seybold *et al.*, 2006; Reid and Purcell, 2011; Manning and Reid,
325 2013), our study expands understanding of the mechanisms by which MPB establish new populations

326 and increase their range; however, due to the heteroscedasticity of the data, morphological variation
327 alone provides limited power to predict flight capacity; many individuals fly short distances despite
328 having the morphology of long distance dispersers. This suggests that, even in a controlled laboratory
329 setting, other intrinsic factors also influence flight and dispersal behaviour. Further research should
330 target these intrinsic influences so that we can better understand the elements governing dispersal-
331 related behaviours of MPB and other economically significant insect species.

332

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344

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1 **Figure Captions**

2 Figure 1. The positive skew in the distribution of flight distances for 124 female mountain pine beetles.
3 Beetles that flew less than 10 m were excluded from this study; the minimum flight distance is 0.01 km.

4

5 Figure 2. Morphological differences between 31 longest distance (> 11.13 km) and 31 shortest distance
6 (< 1.05 km) dispersing female mountain pine beetles. Long distance dispersers have greater preflight
7 and post flight weight, wing size, proportional and absolute weight loss. Points represent group
8 averages and bars show standard error. Comparisons were done using ANOVA. All graphs depict
9 significant results.

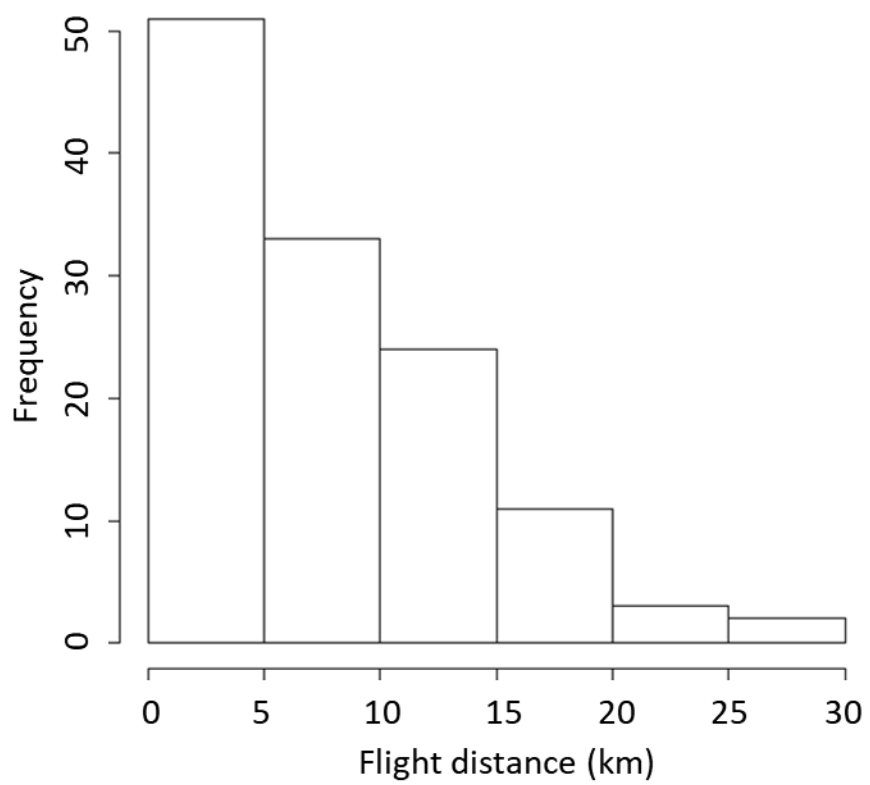
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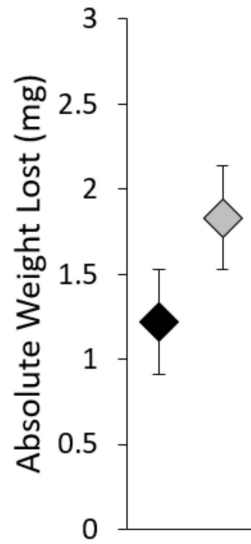
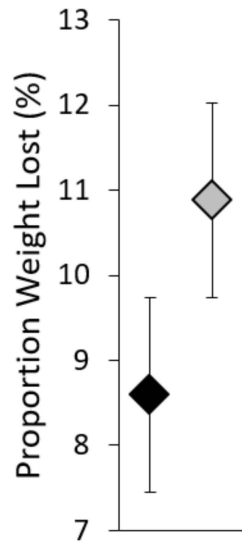
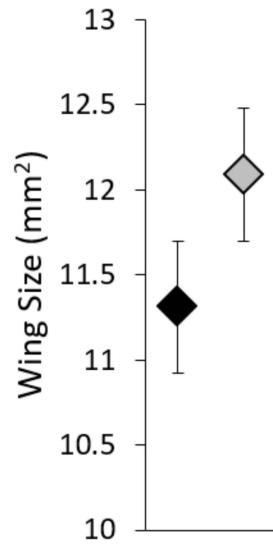
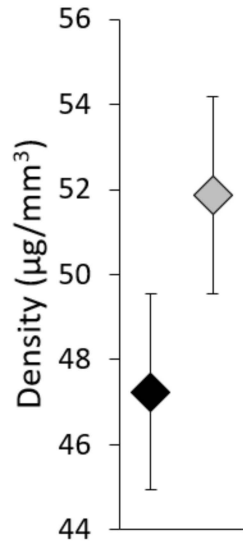
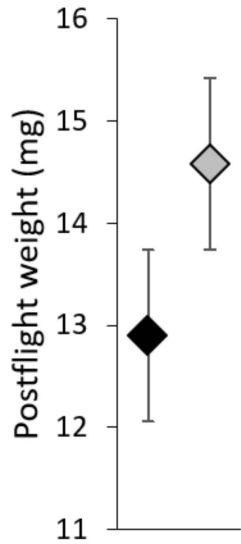
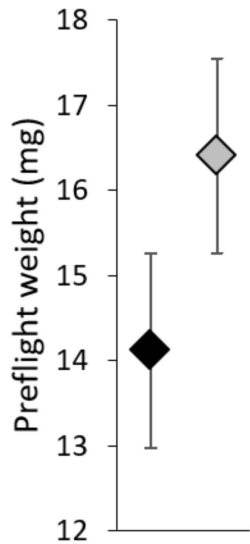
11 Figure 3. Quantile regressions showing the relationships of (A) preflight weight and (B) wing size with
12 flight distance at 12% quantiles of 124 female mountain pine beetles. The solid horizontal line is the
13 average relationship between flight distance and wing loading (dashed line is standard error). Grey
14 shaded area represents the 90 percent confidence band for the quantile estimates. Preflight weight
15 shows a significant deviation from the average relationship at the 96% quantile, showing that higher
16 preflight weight has an increasing effect on flight distance. Although other potential sources of
17 heteroscedasticity may exist, no other significant differences in slope coefficients were found. The
18 asterisk in (A) indicates a significantly different quantile from the average.

19

20 Figure 4. (A) Wing loading vs flight distance of 124 mountain pine beetles, with regression lines of the
21 lowest 12% of dispersers (i) shown as a dotted trendline, and the highest 12% of dispersers (ii) shown as
22 a dashed trendline. (B) Regression at each 12% quantile shows a significant change in the relationship,

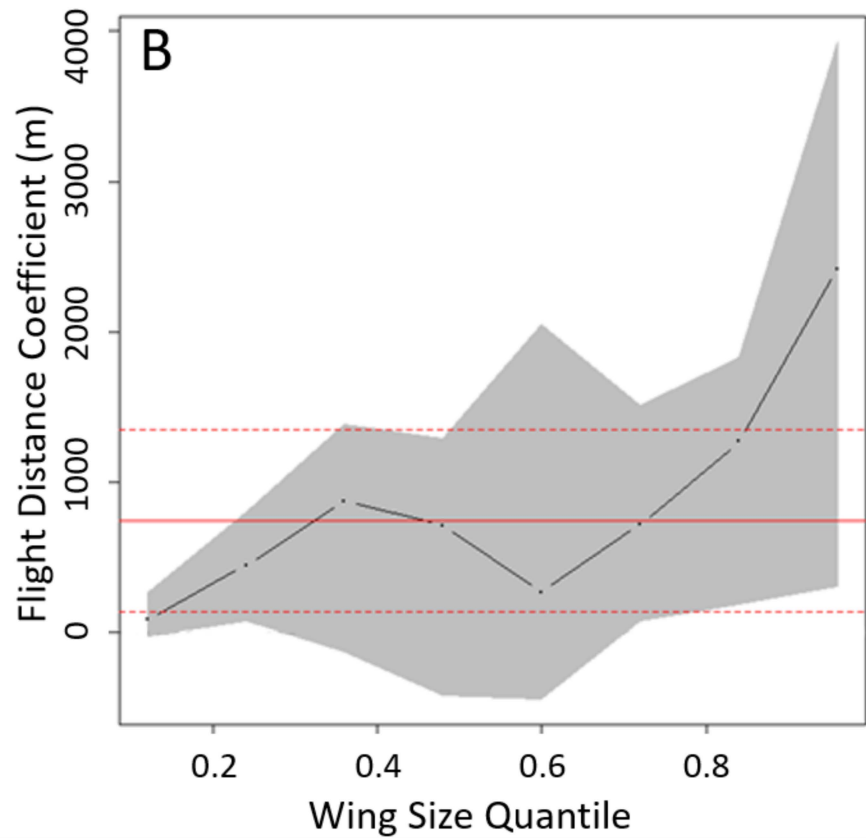
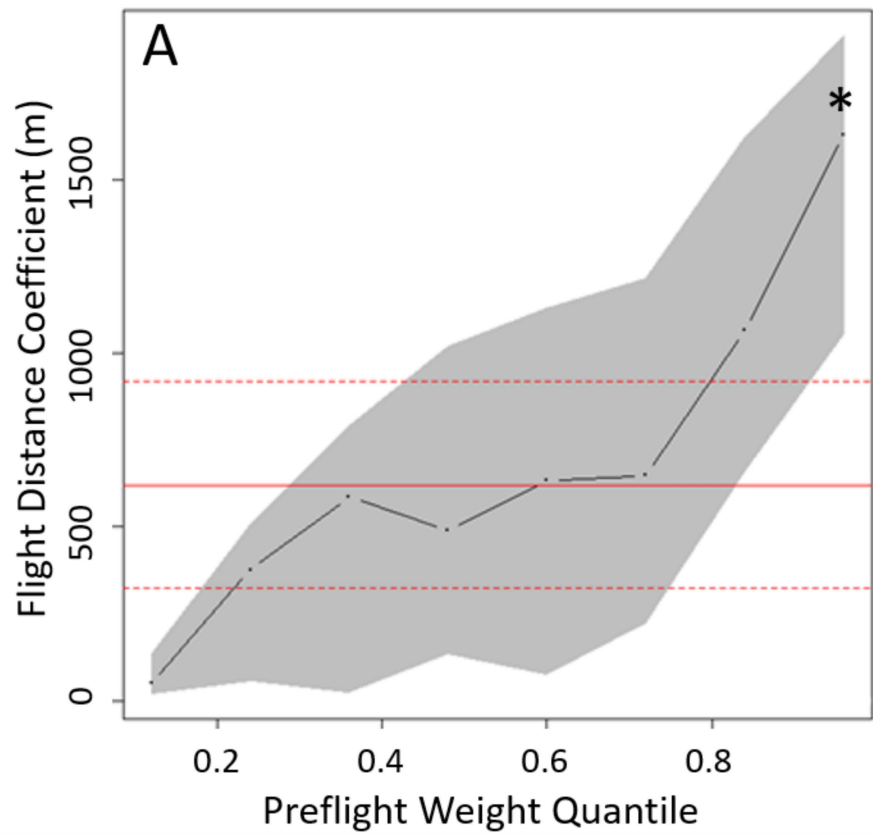
23 resulting in heteroscedasticity. The solid horizontal line is the average relationship between flight
24 distance and wing loading (dashed line is standard error). Grey shaded area represents the 90 percent
25 confidence band for the quantile estimates. At lower quantiles of wing loading (i) the relationship
26 approaches 0; at higher quantiles (ii) the relationship becomes more pronounced. All strong fliers show
27 high wing loading ((A), box 'a'), but many "lazy" individuals have high wing loading and do not fly long
28 distances ((A), box 'b'), causing heteroscedasticity in the relationship. Asterisks in (B) indicate quantiles
29 with significantly different coefficients from the average.

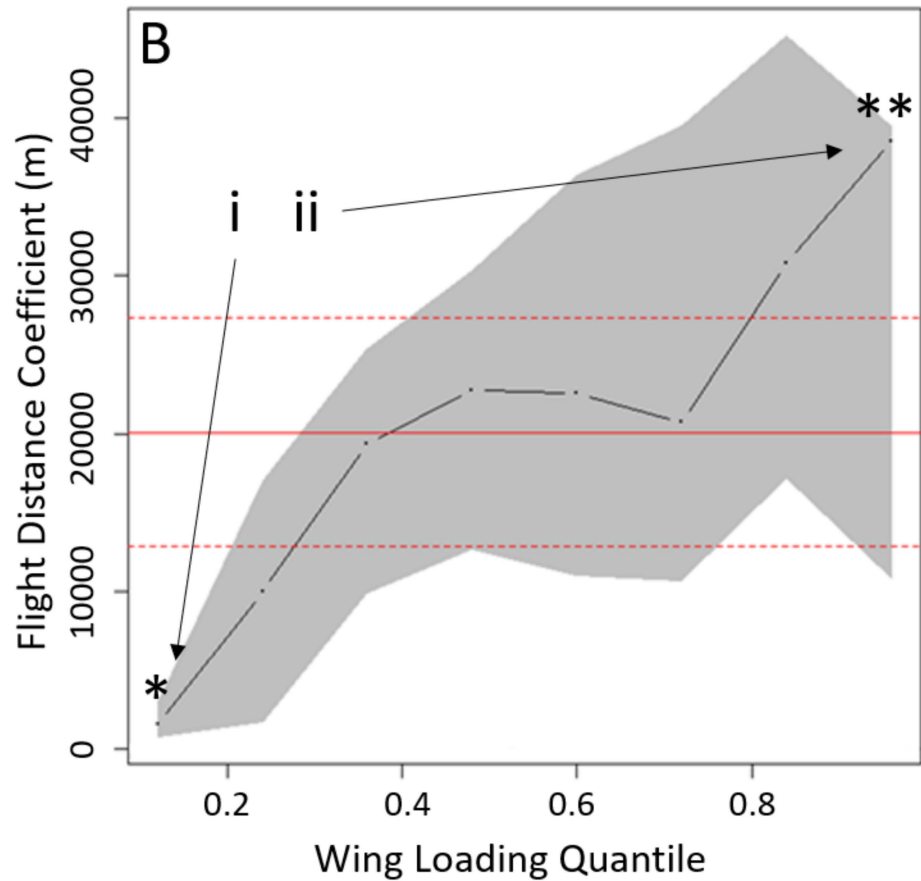
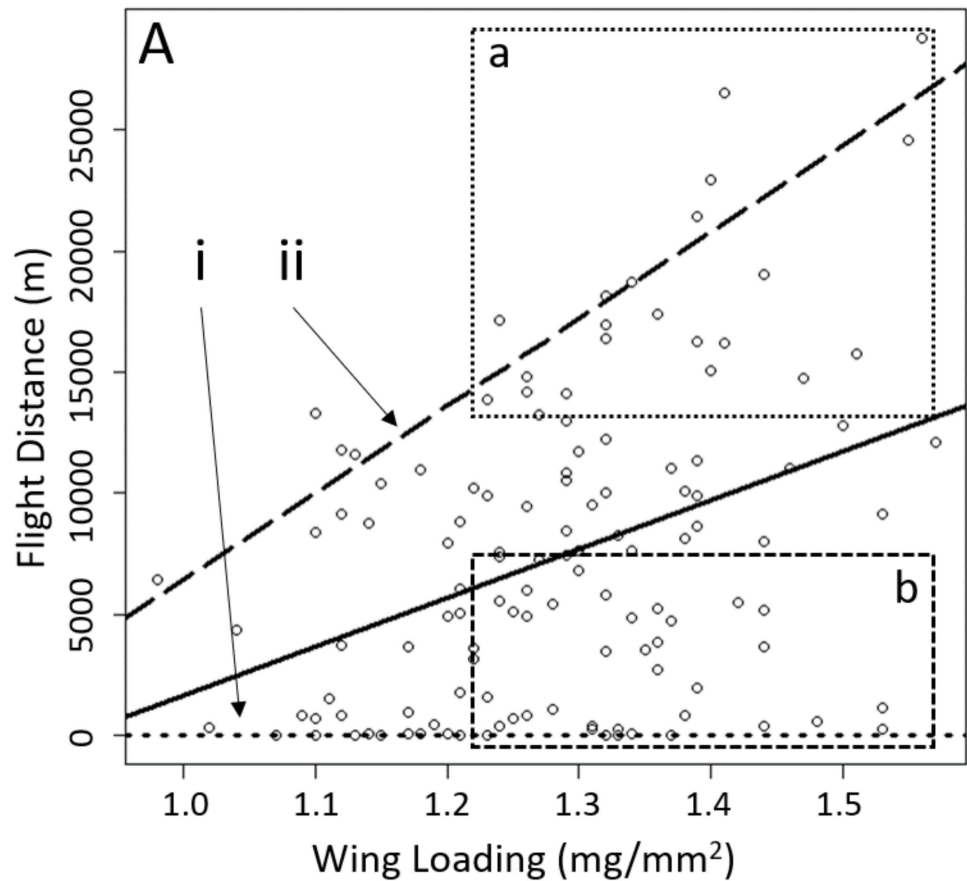


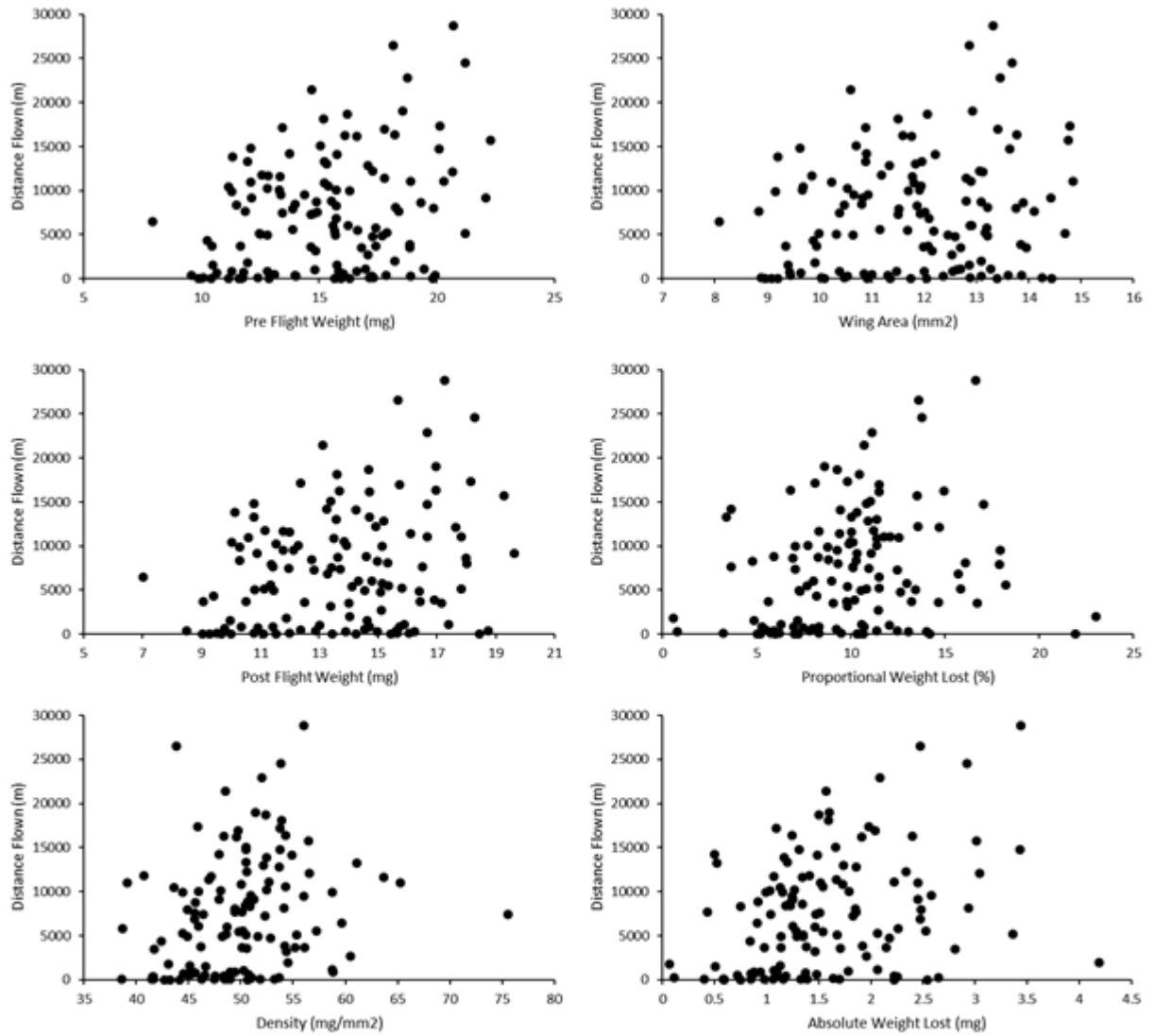


◆ - Short distance flier average

◇ - Long distance flier average







Supplementary Figure S1. Plots showing all 124 specimens relating morphological features to the distance flown in meters, as measured on 22-hour computer-linked flight mill bioassays.