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1 Occupancy patterns of the introduced, predatory sugar glider in Tasmanian forests.

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5

6 **ABSTRACT**

7 Introduced mammals pose serious threats to native island fauna, and understanding their
8 distribution is fundamental to evaluating their conservation impact. Introduced sugar gliders
9 (*Petaurus breviceps*) are the main predator of critically endangered swift parrots (*Lathamus*
10 *discolor*) on mainland Tasmania. We surveyed sugar glider occurrence over ~800 km² in an
11 important swift parrot breeding area, the Southern Forests. During repeated visits to sites we
12 used call broadcast of predatory owls to elicit sugar glider alarm calls and surveyed 100 sites
13 during February/March 2016. Naïve occupancy by sugar gliders was high (0.79), as was
14 detectability (0.523 ± 0.03 s.e.) resulting in a cumulative detection probability of effectively 1.
15 Occupancy modelling indicated a positive effect of the proportion of mature forest cover on
16 occupancy. The best model, based on AIC scores, included the proportion of mature forest
17 cover within a 500m radius with constant detectability. Our study revealed surprisingly high
18 rates of occupancy of available forest habitat throughout the heavily logged study area, such
19 that even when mature forest cover was < 10 %, sugar glider occupancy was > 0.5. Where
20 forest cover approached 100% (i.e. in the best quality breeding habitat for swift parrots),
21 occupancy by sugar gliders approached 1. Our results reveal that sugar gliders are widespread
22 across the study area and may be indicative of occupancy rates elsewhere in the breeding
23 range of the critically endangered swift parrot. As a result, the risk of predation for small
24 birds may be widespread across logged Tasmanian forests. Additional work to identify
25 whether population densities of sugar gliders vary with forest cover (and whether this may

26 impact predation likelihood) is critical to understanding the conservation consequences of
27 deforestation in the breeding range of the swift parrot.

28

29 **Key words**

30 *Petaurus breviceps*, occupancy modelling, predation, swift parrot *Lathamus discolor*,
31 deforestation

32

33 **Introduction**

34 Introduced species threaten global biodiversity (Blackburn *et al.* 2004; Simberloff *et*
35 *al.* 2013). Understanding and addressing the impacts of introduced species on small islands
36 can be straightforward, however, large islands can substantially hinder management actions
37 and knowledge of patterns of occurrence due to the logistic challenges imposed by
38 topography and survey area (Nogales *et al.* 2004; Towns and Broome 2003). On large
39 islands, introduced species that are cryptic or occur at low densities may be difficult to detect,
40 which may limit efficacy of conservation management if action is targeted at suboptimal
41 locations.

42

43 Arboreal nocturnal mammals pose particular challenges for standardised surveys
44 because they are often difficult to detect and can occur at low densities in (often) challenging
45 terrain (Goldingay and Sharpe 2004). Survey methods for arboreal mammals often involve
46 long surveys at night using a range of techniques (e.g. call broadcast, spotlight searches) and
47 imperfect detection (or false absences) is a common problem (Wintle *et al.* 2005). Occupancy
48 modelling (utilising presence/absence data) accounting for imperfect detection is now a
49 commonly used technique to understand species occurrence (MacKenzie *et al.* 2006).

50 Overcoming the problem of false absences often involves a trade-off between the time spent
51 during a single site visit (for example by surveying for longer periods) and spatial replication
52 of the area surveyed. For species that are rare and/or have large potential distributions
53 maximising detectability while minimising the time required for a single site visit can allow
54 far greater spatial replication thus increasing sampling effort and or spatial coverage (Bowler
55 *et al.* 2016; Crates *et al.* 2017; Webb *et al.* 2017).

56

57 Here we use an occupancy modelling framework to identify the distribution of an
58 introduced arboreal marsupial, the sugar glider (*Petaurus breviceps*). Sugar gliders were
59 introduced to Tasmania during the 1830s (Gunn 1851) and unlike in its native range
60 (Lindenmayer 2002), the introduced Tasmanian population is poorly studied (Heinsohn
61 2004). Tasmanian sugar gliders are the main predator of critically endangered swift parrots
62 *Lathamus discolor* and other small cavity nesting birds (Stojanovic *et al.* 2014b). Sugar
63 gliders occur across the swift parrot breeding range, excluding offshore islands, but little is
64 known about their occurrence at finer spatial scales (Heinsohn *et al.* 2015). However, mature
65 forest extent may impact the predatory behaviour of sugar gliders (Stojanovic *et al.* 2014b),
66 but the mechanisms behind this relationship are unknown. We aim to advance knowledge of
67 the occurrence of sugar gliders in the swift parrots breeding range, and examine the potential
68 effect of mature forest cover (and other factors) on glider occurrence. We discuss our results
69 in context of the ecological impact of sugar gliders on the conservation of Tasmanian cavity
70 nesting birds.

71

72 **Methods**

73 *Study area*

74 We surveyed ~800 km² across a key swift parrot breeding area in southern Tasmania,
75 Australia. The Southern Forests are characterized by wet *Eucalyptus globulus*, *E.*
76 *delegatensis*, *E. regnans*, *E. nitida* and *E. obliqua* dominated forests severely fragmented by
77 industrial scale logging. The forests comprise a patchwork of cleared land, regenerating and
78 old-growth native forest and plantation (Hickey 1994). Across much of the study area, the
79 understory is dominated by temperate rainforest and other mesic vegetation. Mean minimum
80 and maximum temperatures in the region range from 10 - 22°C in February to 2 - 12°C in
81 July and average annual rainfall is 877mm (BOM 2016). Elevation of the survey sites ranged
82 from 12 to 687 m.

83

84 *Study Design*

85 We selected 100 survey sites over the study area (Figure 1) including existing swift
86 parrot monitoring sites (Webb *et al.* 2017; Webb *et al.* 2014) and additional sites selected
87 using the following criteria. All sites contained at least one mature, cavity-bearing *Eucalyptus*
88 within 100 m of the centroid (i.e. potential sugar glider/swift parrot habitat) and were at least
89 500 m away from other sites. Sites were defined as a 200 m radius around the centroid.
90 Repeated five minute site visits (4 - 5 visits/site) were undertaken during February/March of
91 2016. Based on the results of a pilot study (Allen, M. unpublished data) we improved sugar
92 glider detectability using southern boobook *Ninox novaeseelandiae* call broadcast to elicit
93 alarm calls from sugar gliders. Surveys consisted of two minutes of listening, followed by
94 three minutes of intermittent call broadcast from a portable speaker.

95

96 To reduce the potential effects of weather on sugar glider detectability (sugar gliders
97 become torpid during inclement weather (Lindenmayer 2002)), surveys were only conducted
98 between 21:00-02:00 h, within a temperature range of 10 – 20°C and when wind speeds were

99 < 20 km/h (i.e. clement weather). A fifth survey was undertaken at sites where the gliders
100 were not detected in the first four surveys. During surveys we recorded: (i) sugar glider
101 detection/non-detection ; (ii) wind speed; (iii) temperature, scored as: 1 = 9 - 12 ° C, 2 = 13 -
102 16 ° C, 3 = 17 - 20 ° C; (iv) moon brightness, scored as: 0= new moon, 1 = small crescent
103 moon, 2 = large crescent moon, 3 = full moon, and; (v) southern boobook detection/non-
104 detection (based on calls). Temperature and wind speed were measured with a Kestrel 3000
105 RH/Wind Meter (Nielson-Kellerman, Boothwyn, PA, USA).

106 We used ArcMap 10.3 to derive site-level variables: (i) elevation and (ii) mature
107 forest cover within the following radii from the site: 200 m, 500 m, 1000 m, 1500 m and
108 2000 m. Mature forest cover was estimated using the aerial forest inventories that quantify
109 the spatial extent of mature, cavity-bearing forest (FPA 2011), and are a good indication of
110 potential habitat for tree cavity dependent animals (Stojanovic *et al.* 2014a). We followed
111 (Stojanovic *et al.* 2014b) and pooled data for three categories (low, medium and high) of
112 mature forest cover because all constitute potential sugar glider habitat because of the
113 occurrence of mature trees in each.

114

115 *Statistical analysis*

116 To quantify patterns of sugar glider occurrence across the study area we followed an
117 occupancy modelling approach (MacKenzie *et al.* 2006), using single-season models
118 implemented in R (R Core Team 2017) using the package *unmarked* (Fiske and Chandler
119 2011). We modelled sugar glider occupancy and detectability and included site-level (forest
120 covers, elevation) covariates that could impact occupancy, and observation-level
121 (temperature, moon phase, owl occurrence) covariates that could impact detection. We
122 selected the best model using the Akiake Information Criterion and we tested goodness of fit
123 using the parametric bootstrap method with 1000 simulations.

124

125 **Results**

126 Naïve occupancy (i.e. the proportion of sites sugar gliders were detected) was 0.79,
127 almost equivalent to the modelled occupancy estimate of 0.81 assuming constant occupancy
128 and detection. Twenty four sites were visited a fifth time because gliders had not been
129 detected in the first four visits (total 424 surveys). Because we controlled for the potential
130 effect of wind, 83 % of site visits had wind speeds < 5 km/h and all were < 10 km/h. Due to
131 this small variation, we excluded wind speed from further analysis. The best model, based on
132 AIC scores (Table 1) included a significant included a significant positive effect of mature
133 forest within 500 m of the site on likelihood of sugar glider occupancy (estimate: 5.51 ± 2.54 ,
134 z : 2.17, p : 0.03) and a non-significant effect of temperature in the detectability component (p :
135 0.1). The second best model based on AIC also included a positive effect of mature forest
136 within 500 m of the site on likelihood of sugar glider occupancy (estimate: 5.58 ± 2.48 , z :
137 2.25, p : 0.02) and assumed constant detectability. We selected the simpler (constant
138 detectability) model as our preferred (Figure 2). Given estimated detectability for the best
139 model was 0.523 (± 0.03 SE), the cumulative probability of detecting sugar gliders if they
140 were present at a site, was 95 % by the fourth site visit (Figure 3).

141

142 **Discussion**

143 We used an occupancy framework to undertake a rapid, landscape scale survey of
144 sugar gliders within a key breeding area for swift parrots. Our results reveal high rates of
145 sugar glider occupancy across the study area, and a positive correlation with the proportion of
146 mature forest cover within 500 m radius. Even when mature forest cover was low (< 10 %),
147 sugar glider site occupancy of survey sites was still greater than 0.5. This finding underscores
148 the widespread predation risk for small cavity nesting birds in this landscape even in small

149 habitat fragments. Sugar glider predation on birds may be correlated with forest disturbance,
150 such that areas of low forest cover suffer the worst predation rates (Stojanovic *et al.* 2014b).
151 In their native range, sugar gliders are common in fragmented landscapes (Suckling 1984)
152 and are known to tolerate logging (Kavanagh and Bamkin 1995).

153 We demonstrate the efficacy of short surveys incorporating predatory owl call
154 broadcast for surveying sugar gliders. Sugar gliders can be challenging to detect using
155 spotlighting searches because they commonly turn away from lights, and have relatively dull
156 retinal reflectance compared to other nocturnal mammals (Wintle *et al.* 2005). We had a 77
157 % likelihood of detecting sugar gliders with only two five minute site visits, and this
158 likelihood increased to 89 % and 95 % for three and four visits respectively (Figure 3). The
159 method we used is fast and low cost, facilitating increased spatial replication across our large
160 study area. Controlling for survey conditions (i.e. good weather) improved survey efficacy
161 even in challenging terrain because sugar gliders may be heard calling over hundreds of
162 meters on calm nights. This approach was particularly valuable in our study area, where
163 access to off-road study sites can be challenging even in daylight hours.

164 Our results have serious implications for swift parrots and other cavity nesting birds
165 because sugar gliders are resident in their territories, and occupy most of the potential swift
166 parrot habitat in the study area. Areas with greater abundance of mature forest may be
167 attractive for the group-nesting swift parrot, but depending on fine scale habitat
168 configuration, nests may suffer severe predation (Heinsohn *et al.* 2015). More data on
169 abundance and behavior of sugar gliders both at large (swift parrot breeding range) and fine
170 scales (sugar glider home ranges) is necessary to identify the mechanisms that underpin the
171 relationship between forest cover and nest predation. Fine scale variation in population
172 densities of sugar gliders may have important ramifications for bird nesting success, and

173 given the high conservation threat sugar gliders pose to Tasmanian birds, this warrants urgent
174 attention.

175 We demonstrate the conservation value of identifying efficient survey approaches for
176 invasive species to overcome the challenges of monitoring large areas of rugged terrain.
177 Given the vulnerability of island species to introduced predators, overcoming data limitations
178 about where predators occur is a critical first step to conserving vulnerable native species.
179 Our study demonstrates an effective approach to detecting potential predation risk that, in
180 combination with information about where swift parrots are likely to nest, provides a useful
181 management tool for prioritising areas for nest protection.

182

183 **Acknowledgements**

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185 Australian Government National Environmental Science Programme via the Threatened
186 Species Research Hub. The research was conducted under approval from the Australian
187 National University Animal Ethics Committee (A2014/26) and under permit from the
188 Tasmanian Government (TFA16234).

189

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257 **Tables and Figures**

258

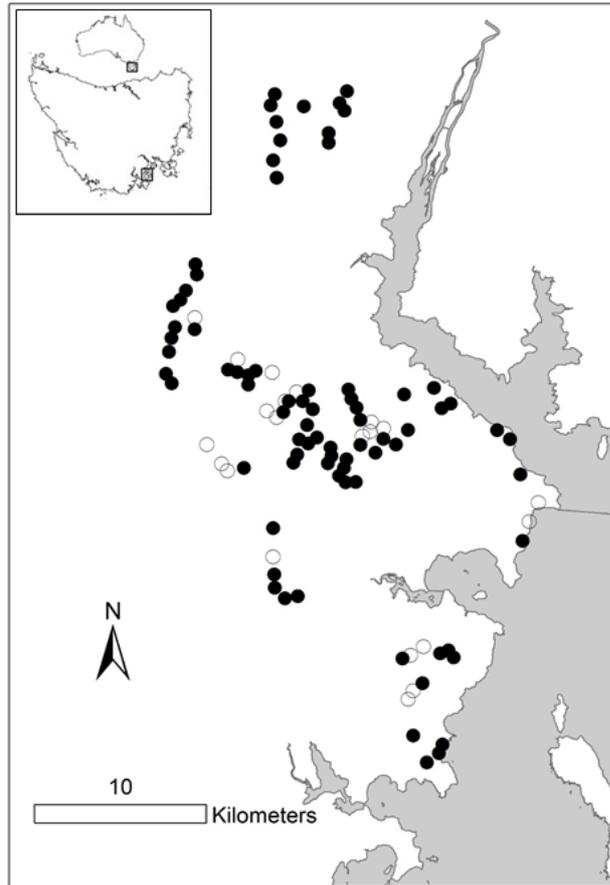
259 **Table 1.** Candidate models of sugar glider occupancy (Ψ) and detectability (p). nPars =

260 number of parameters.

Model	nPars	AIC
$\Psi(500m) . p(\text{temperature})$	4	534.32
$\Psi(500m) . p(.)$	3	534.99
$\Psi(500m) . p(\text{moon})$	4	536.38
$\Psi(500m) . p(\text{owl})$	4	536.99
$\Psi(200m) . p(\text{temperature})$	4	538.74
$\Psi(200m) . p(.)$	3	539.39
$\Psi(1000m) . p(\text{temperature})$	4	539.77
$\Psi(1000m) . p(.)$	3	540.6
$\Psi(200m) . p(\text{moon})$	4	540.9
$\Psi(200m) . p(\text{owl})$	4	541.38
$\Psi(1000m) . p(\text{moon})$	4	542.01
$\Psi(.) . p(\text{temperature})$	3	542.42
$\Psi(1000m) . p(\text{owl})$	4	542.58
$\Psi(1500m) . p(\text{temperature})$	4	543.03
$\Psi(.) . p(.)$	2	543.25
$\Psi(1500m) . p(.)$	3	543.82
$\Psi(2000m) . p(\text{temperature})$	4	544.13
$\Psi(\text{elevation}) . p(\text{temperature})$	4	544.35
$\Psi(.) . p(\text{moon})$	3	544.75
$\Psi(2000m) . p(.)$	3	544.95
$\Psi(\text{elevation}) . p(.)$	3	545.16
$\Psi(\text{elevation}) . p(\text{moon})$	3	545.23
$\Psi(.) . p(\text{owl})$	3	545.23
$\Psi(1500m) . p(\text{moon})$	4	545.27
$\Psi(1500m) . p(\text{owl})$	4	545.8

$\Psi(2000m) \cdot p(\text{moon})$	4	546.43
$\Psi(2000m) \cdot p(\text{owl})$	4	546.93
$\Psi(\text{elevation}) \cdot p(\text{owl})$	4	547.12

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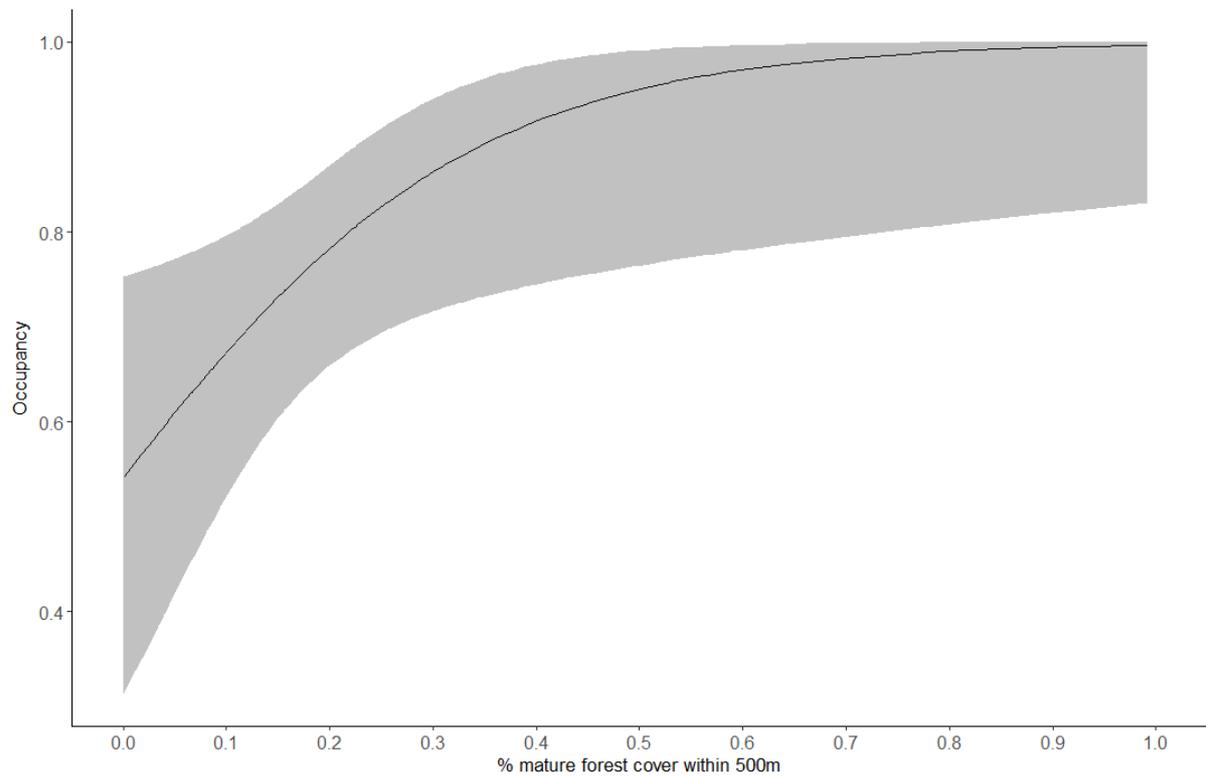
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263 **Figure 1.** Study area showing the location of the survey sites where sugar gliders were
 264 present (black) and absent (white).

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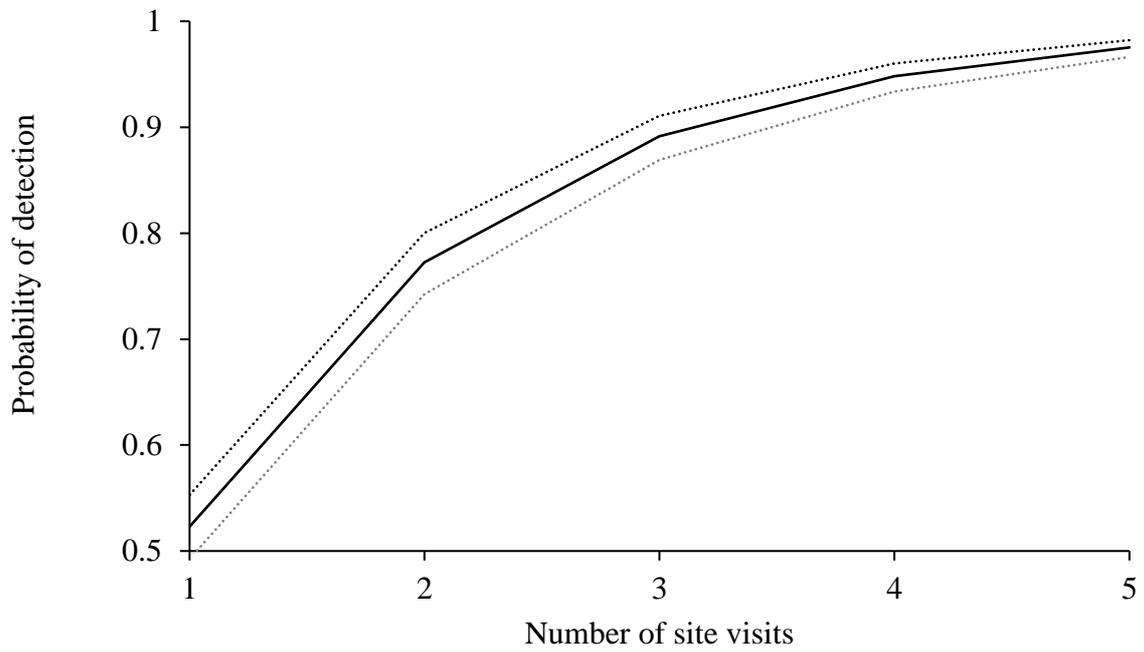


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269 **Figure 2.** Modelled probability of sugar glider site occupancy relative to cover of mature
270 forest within 500 m of the survey site centroid. Line is the occupancy predicted and grey area
271 represents 95% confidence interval.

272

273



275

276 **Figure 3.** Cumulative detection probability of sugar gliders. The black line represents a
277 detection probability of 0.523 (± 0.03 s.e.) for a single site visit, derived from the best model.

278

279