

Supplementary material

1 Appendix 1

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3 Table A1. Model selection for Aride Island's two-species shearwater samplings based on the
 4 Deviance Information Criterion (DIC). (z) denotes plot variation in the detection parameter,
 5 while (.) denotes constant detection rates for all plots.

Model		DIC	Δ DIC
<i>P.pacificus</i>	<i>P.bailloni</i>		
$p_R(z), p_S(z)$	$p_R(z), p_S(z)$	627.51	0.00
$p_R(z), p_S(z)$	$p_R(z), p_S(.)$	651.59	24.08
$p_R(.), p_S(z)$	$p_R(.), p_S(z)$	654.40	26.89
$p_R(z), p_S(.)$	$p_R(z), p_S(.)$	657.09	29.58
$p_R(z), p_S(z)$	$p_R(.), p_S(z)$	662.37	34.86
$p_R(z), p_S(.)$	$p_R(z), p_S(z)$	665.58	38.07
$p_R(z), p_S(.)$	$p_R(.), p_S(.)$	666.35	38.84
$p_R(.), p_S(z)$	$p_R(z), p_S(z)$	666.66	39.14
$p_R(.), p_S(z)$	$p_R(.), p_S(.)$	682.89	55.38
$p_R(z), p_S(z)$	$p_R(.), p_S(.)$	684.58	57.07
$p_R(.), p_S(.)$	$p_R(.), p_S(z)$	684.93	57.42
$p_R(z), p_S(.)$	$p_R(.), p_S(.)$	686.40	58.89
$p_R(.), p_S(.)$	$p_R(z), p_S(z)$	688.23	60.71
$p_R(.), p_S(z)$	$p_R(z), p_S(.)$	695.90	68.39
$p_R(.), p_S(.)$	$p_R(z), p_S(.)$	702.62	75.11
$p_R(.), p_S(.)$	$p_R(.), p_S(.)$	713.44	85.93

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8 Table A2. Parameter estimates for Aride Island's two-species shearwater sampling. The best
 9 model includes variation in occupancy, sighting and response probabilities for both species

Parameter	Mean Estimate	SD
Wedge-tailed Shearwater		
μ_{ψ}	-1.224	0.471
σ_{ψ}	1.109	0.534
μ_{ps}	0.948	0.349
σ_{ps}	0.909	0.615
μ_{pr}	2.011	1.067
σ_{pr}	1.656	1.620
Tropical Shearwater		
μ_{ψ}	-0.493	0.869
σ_{ψ}	3.107	1.184
μ_{ps}	0.201	9.875
σ_{ps}	4.994	2.890
μ_{pr}	4.431	0.384
σ_{pr}	0.754	0.471

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11

12 Table A3. Root mean squared error, bias and percentage of realized number of occupied
 13 burrows falling in the estimated 95% confidence intervals for different combinations of
 14 parameters in the simulations with two species

Parameters						RMSE		Bias		Proportion of realised values in the estimated confidence intervals	
ψ_A	ψ_B	P_{VA}	P_{RA}	P_{VB}	P_{RB}	species A	species B	species A	species B	species A	species B
0.3	0.3	0.3	0.3	0.3	0.3	1.83	1.45	0.41	0.38	0.9	0.95
0.3	0.3	0.3	0.5	0.3	0.3	1.69	1.42	0.52	0.32	0.9	0.95
0.3	0.3	0.3	0.5	0.3	0.5	1.26	2.15	0.27	0.82	0.7	0.8
0.3	0.3	0.3	0.7	0.3	0.3	2.26	0.62	0.71	0.13	0.95	0.9
0.3	0.3	0.3	0.7	0.3	0.5	1.77	0.73	0.45	0.10	0.75	0.75
0.3	0.3	0.3	0.7	0.3	0.7	1.07	1.26	0.17	0.27	0.85	0.7
0.3	0.3	0.5	0.3	0.3	0.3	1.38	1.43	0.25	0.16	1	0.85
0.3	0.3	0.5	0.3	0.5	0.3	1.59	1.44	0.51	0.25	0.85	0.85
0.3	0.3	0.5	0.5	0.3	0.3	1.20	1.22	0.19	0.43	0.95	0.8
0.3	0.3	0.5	0.5	0.3	0.5	1.89	1.51	0.39	0.36	0.9	0.7
0.3	0.3	0.5	0.5	0.5	0.3	1.53	1.09	0.45	0.45	0.9	0.9
0.3	0.3	0.5	0.5	0.5	0.5	1.07	1.26	0.29	0.38	0.75	0.95
0.3	0.3	0.5	0.7	0.3	0.3	1.47	0.86	0.28	0.23	0.95	0.95
0.3	0.3	0.5	0.7	0.3	0.5	1.48	1.00	0.22	0.26	0.75	0.65
0.3	0.3	0.5	0.7	0.3	0.7	1.10	0.91	0.34	0.22	0.9	0.8
0.3	0.3	0.5	0.7	0.5	0.3	1.68	0.83	0.51	0.22	0.9	0.65
0.3	0.3	0.5	0.7	0.5	0.5	1.39	0.80	0.45	0.19	0.8	0.7
0.3	0.3	0.5	0.7	0.5	0.7	1.36	0.75	0.39	0.13	0.8	0.7
0.3	0.3	0.7	0.3	0.3	0.3	1.77	1.31	0.25	0.34	0.8	0.85
0.3	0.3	0.7	0.3	0.5	0.3	1.61	0.85	0.47	0.22	0.95	0.65
0.3	0.3	0.7	0.3	0.7	0.3	0.77	1.11	0.22	0.18	0.9	0.9
0.3	0.3	0.7	0.5	0.3	0.3	1.39	0.95	0.21	0.24	0.9	0.75
0.3	0.3	0.7	0.5	0.3	0.5	1.31	1.27	0.41	0.46	1	0.75
0.3	0.3	0.7	0.5	0.5	0.3	1.47	1.14	0.32	0.21	0.85	0.6
0.3	0.3	0.7	0.5	0.5	0.5	1.10	1.21	0.22	0.46	0.7	0.85
0.3	0.3	0.7	0.5	0.7	0.3	1.17	1.03	0.31	0.23	0.8	0.85
0.3	0.3	0.7	0.5	0.7	0.5	0.75	0.72	0.15	0.08	0.85	0.7
0.3	0.3	0.7	0.7	0.3	0.3	1.37	0.60	0.36	0.14	0.9	0.5
0.3	0.3	0.7	0.7	0.3	0.5	1.56	0.64	0.39	0.16	0.85	0.6
0.3	0.3	0.7	0.7	0.3	0.7	1.01	0.57	0.25	0.10	0.9	0.7
0.3	0.3	0.7	0.7	0.5	0.3	1.19	0.53	0.30	0.01	0.8	0.6
0.3	0.3	0.7	0.7	0.5	0.5	1.14	0.91	0.28	0.29	0.85	0.65
0.3	0.3	0.7	0.7	0.5	0.7	0.87	0.85	0.08	0.24	0.75	0.5
0.3	0.3	0.7	0.7	0.7	0.3	1.11	0.35	0.32	0.05	0.85	0.5
0.3	0.3	0.7	0.7	0.7	0.5	0.82	0.91	0.28	0.30	0.75	0.5
0.3	0.3	0.7	0.7	0.7	0.7	0.37	0.81	-0.01	0.12	0.85	0.6
0.5	0.3	0.3	0.3	0.3	0.3	0.88	1.14	0.07	0.15	0.95	0.95
0.5	0.3	0.3	0.5	0.3	0.3	0.49	1.10	0.04	0.13	1	1

Parameters						RMSE		Bias		Proportion of realised values in the estimated confidence intervals	
Ψ_A	Ψ_B	P_{VA}	P_{RA}	P_{VB}	P_{RB}	species A	species B	species A	species B	species A	species B
0.5	0.3	0.3	0.5	0.3	0.5	0.72	1.39	0.17	0.17	0.95	1
0.5	0.3	0.3	0.7	0.3	0.3	0.80	0.96	0.14	0.12	0.9	0.95
0.5	0.3	0.3	0.7	0.3	0.5	0.88	0.97	0.36	0.11	0.95	0.95
0.5	0.3	0.3	0.7	0.3	0.7	0.57	1.08	0.02	0.04	0.95	0.95
0.5	0.3	0.5	0.3	0.3	0.3	0.68	1.16	0.09	0.14	1	0.95
0.5	0.3	0.5	0.3	0.5	0.3	0.62	1.19	0.19	0.15	1	1
0.5	0.3	0.5	0.5	0.3	0.3	1.00	1.14	0.13	0.10	0.85	0.95
0.5	0.3	0.5	0.5	0.3	0.5	0.84	1.03	0.28	0.09	1	0.95
0.5	0.3	0.5	0.5	0.5	0.3	0.83	1.08	0.11	0.12	0.85	0.9
0.5	0.3	0.5	0.5	0.5	0.5	0.59	1.03	0.14	0.07	1	0.95
0.5	0.3	0.5	0.7	0.3	0.3	0.76	0.59	0.07	0.04	1	1
0.5	0.3	0.5	0.7	0.3	0.5	0.59	1.00	0.08	0.20	0.95	0.95
0.5	0.3	0.5	0.7	0.3	0.7	0.54	0.87	0.15	0.07	0.9	1
0.5	0.3	0.5	0.7	0.5	0.3	0.71	0.95	0.20	0.11	1	1
0.5	0.3	0.5	0.7	0.5	0.5	0.48	0.77	0.09	0.13	0.95	0.95
0.5	0.3	0.5	0.7	0.5	0.7	0.43	0.62	0.15	0.02	1	1
0.5	0.3	0.7	0.3	0.3	0.3	0.76	1.18	0.02	0.20	0.9	0.95
0.5	0.3	0.7	0.3	0.5	0.3	0.77	1.01	0.22	0.13	0.95	1
0.5	0.3	0.7	0.3	0.7	0.3	0.52	0.69	0.10	0.08	1	1
0.5	0.3	0.7	0.5	0.3	0.3	0.94	0.61	0.23	0.02	0.85	0.95
0.5	0.3	0.7	0.5	0.3	0.5	0.64	1.13	0.04	0.20	0.95	0.85
0.5	0.3	0.7	0.5	0.5	0.3	0.63	1.10	0.08	0.19	1	0.95
0.5	0.3	0.7	0.5	0.5	0.5	0.65	0.52	0.19	0.08	0.95	0.95
0.5	0.3	0.7	0.5	0.7	0.3	0.58	0.90	0.07	0.09	0.95	0.95
0.5	0.3	0.7	0.5	0.7	0.5	0.43	0.67	0.12	0.10	1	0.9
0.5	0.3	0.7	0.7	0.3	0.3	0.73	0.84	-0.01	0.12	0.85	0.85
0.5	0.3	0.7	0.7	0.3	0.5	0.95	0.74	0.29	0.05	0.95	1
0.5	0.3	0.7	0.7	0.3	0.7	0.50	0.84	0.10	0.10	1	0.95
0.5	0.3	0.7	0.7	0.5	0.3	0.73	0.65	0.07	0.07	0.95	0.9
0.5	0.3	0.7	0.7	0.5	0.5	0.50	0.75	0.13	0.08	1	0.9
0.5	0.3	0.7	0.7	0.5	0.7	0.64	0.69	0.25	0.09	1	1
0.5	0.3	0.7	0.7	0.7	0.3	0.65	0.29	0.23	0.04	1	1
0.5	0.3	0.7	0.7	0.7	0.5	0.41	0.62	0.11	0.08	0.95	0.95
0.5	0.3	0.7	0.7	0.7	0.7	0.46	0.54	0.12	-0.02	0.85	0.95

16 **JAGS model for the analysis of the two-species dataset.**

```
17 model{
18     oc.mean.1~dnorm(0,0.01)
19     oc.sigma.1~dunif(0,10)
20     oc.tau.1 <- 1/(oc.sigma.1^2)
21
22     oc.mean.2~dnorm(0,0.01)
23     oc.sigma.2~dunif(0,10)
24     oc.tau.2 <- 1/(oc.sigma.2^2)
25
26     pr.mean.1~dnorm(0,0.01)
27     pr.sigma.1~dunif(0,10)
28     pr.tau.1 <- 1/(pr.sigma.1^2)
29
30     pr.mean.2~dnorm(0,0.01)
31     pr.sigma.2~dunif(0,10)
32     pr.tau.2 <- 1/(pr.sigma.2^2)
33
34     ps.mean.1~dnorm(0,0.01)
35     ps.sigma.1~dunif(0,10)
36     ps.tau.1 <- 1/(ps.sigma.1^2)
37
38     ps.mean.2~dnorm(0,0.01)
39     ps.sigma.2~dunif(0,10)
40     ps.tau.2 <- 1/(ps.sigma.2^2)
```

```

41
42
43     for (j in 1:nquad) {
44         logit_poc.1[j]~dnorm(oc.mean.1,oc.tau.1)
45         poc.1[j]<-
46 exp(logit_poc.1[j])/(1+exp(logit_poc.1[j])+exp(logit_poc.2[j]))
47 )
48
49         logit_poc.2[j]~dnorm(oc.mean.2,oc.tau.2)
50         poc.2[j]<-
51 exp(logit_poc.2[j])/(1+exp(logit_poc.1[j])+exp(logit_poc.2[j]))
52 )
53
54         logit_pr.1[j]~dnorm(pr.mean.1,pr.tau.1)
55         pr.1[j] <-
56 exp(logit_pr.1[j])/(1+exp(logit_pr.1[j]))
57
58         logit_ps.1[j]~dnorm(ps.mean.1,ps.tau.1)
59         ps.1[j] <-
60 exp(logit_ps.1[j])/(1+exp(logit_ps.1[j]))
61
62         logit_pr.2[j]~dnorm(pr.mean.2,pr.tau.2)
63         pr.2[j] <-
64 exp(logit_pr.2[j])/(1+exp(logit_pr.2[j]))
65
66         logit_ps.2[j]~dnorm(ps.mean.1,ps.tau.1)

```

```

67         ps.2[j] <-
68 exp(logit_ps.2[j])/(1+exp(logit_ps.2[j]))
69
70         pi[j,1] <- 1-(poc.1[j] + poc.2[j])
71         pi[j,2] <- poc.1[j]
72         pi[j,3] <- poc.2[j]
73     }
74
75     for (i in 1:ndata){
76         oc[i] ~ dcat(pi[quadrat[i],])
77         occupied[i] <- oc[i] - 1
78
79         he[i,1] ~
80 dbern(equals(occupied[i],1)*pr.1[quadrat[i]])
81         he[i,2] ~ dbern(equals(occupied[i],2)*pr.2[quadrat[i]])
82
83         se[i,1] ~
84 dbern(equals(occupied[i],1)*ps.1[quadrat[i]])
85         se[i,2] ~
86 dbern(equals(occupied[i],2)*ps.2[quadrat[i]])
87
88         occ.1[i] <- equals(occupied[i],1)
89         occ.2[i] <- equals(occupied[i],2)
90     }
91

```



```
92  N.1 <- 5110*sum(occ.1[1:ndata])/nquad
93  N.2 <- 5110*sum(occ.2[1:ndata])/nquad

94  for (q in 1:nquad){
95      for (b in 1:ndata){
96          is.2[b,q]<-equals(quadrat[b],q)*occ.2[b]
97          is.1[b,q]<-equals(quadrat[b],q)*occ.1[b]
98      }
99
100     n.2[q]<-sum(is.2[,q])
101     n.1[q]<-sum(is.1[,q])
102 }

103 }

104

105
```