

Supplementary information

Land management strategies can increase oil palm plantation use by some terrestrial mammals in Colombia

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Fig S1. Detection probability(\hat{p})for most common species across oil palm dominated landscapes in Colombian Llanos. Error bars indicate confidence intervals.

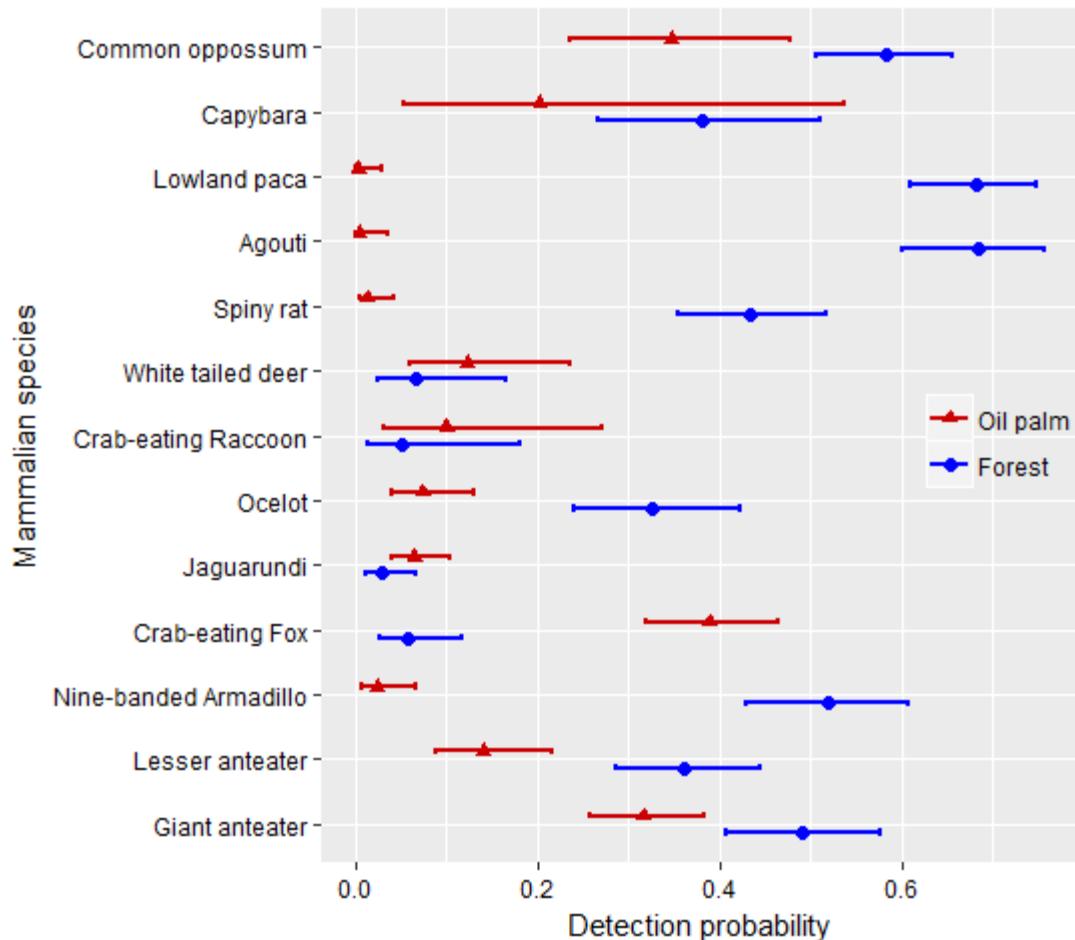


Table S1. Model selection ($\Delta \text{AICc} < 2$) evaluating the effect of habitat type (hab) on habitat use probabilities (Ψ) and detection probabilities (p) for selected mammal species*. Habitat type is a binary variable where 0=forest (intercept) and 1= oil palm plantation (β), a negative value of the corresponding β coefficient suggest preference for forest and vice versa.

Model	AIC	ΔAICc	$\text{AIC } w$	k	β (SE) Oil palm
Giant anteater					
$\Psi(.), p(\text{hab})$	526.26	0	0.501	3	
$\Psi(\text{hab}), p(\text{hab})$	526.32	0.06	0.4862	4	2.12 (2.60)
Lesser anteater					
$\Psi(\text{hab}), p(\text{hab})$	391.21	0	0.4979	4	-1.64 (0.97)
$\Psi(.), p(\text{hab})$	391.32	0.11	0.4713	3	
Nine-banded armadillo					
$\Psi(.), p(\text{hab})$	242.34	0	0.6103	3	
$\Psi(\text{hab}), p(\text{hab})$	243.25	0.91	0.3872	4	-2.31 (1.15)
Naked-tailed armadillo					
$\Psi(\text{hab}), p(.)$	110.55	0	0.7612	3	-26.48 (#)
Crab-eating fox					
$\Psi(.), p(\text{hab})$	365.87	0	0.9965	3	
Jaguarundi					
$\Psi(.), p(.)$	170.78	0	0.5548	2	
$\Psi(\text{hab}), p(.)$	171.22	0.44	0.4452	3	27.03 (#)
Ocelot					
$\Psi(.), p(\text{hab})$	294.61	0	0.9998	3	
Crab-eating raccoon					
$\Psi(.), p(.)$	124.6	0	0.4516	2	
$\Psi(.), p(\text{hab})$	125.76	1.16	0.2529	3	
White-tailed deer					
$\Psi(.), p(.)$	190.76	0	0.3984	2	
$\Psi(.), p(\text{hab})$	191.27	0.51	0.3087	3	
$\Psi(\text{hab}), p(.)$	192.25	1.49	0.1891	3	0.79 (0.95)
Capybara					
$\Psi(\text{hab}), p(.)$	166.95	0	0.5784	3	-1.69 (0.95)
$\Psi(\text{hab}), p(\text{hab})$	168.66	1.71	0.246	4	-1.61 (0.96)
Spiny rat					
$\Psi(\text{hab}), p(\text{hab})$	260.92	0	0.5125	4	-4.20 (1.22)
$\Psi(.), p(\text{hab})$	261.59	0.67	0.3666	3	
Common opossum					
$\Psi(\text{hab}), p(\text{hab})$	404.11	0	0.969	4	-3.62 (1.10)

Notes: ΔAICc = difference in AIC values between each model with the lowest AIC model (best model); $\text{AIC}\omega$ = Akaike weight.; k = number of parameters in the model; SE: standard error. Habitat type, a binary covariate with 0 =

forest (intercept) and 1= oil palm (beta). # = high standard error, so species-specific occupancy estimates are imprecise. However, direction of the effect is not affected (Hines et al 2006). Only species with sufficient data to conduct modeling are shown (i.e. at least 4 detections per habitat). Refer to naïve occupancy in Table 1 for rare species and scientific names.