Supplementary Material

Making inferences about non-detection observations to

improve occurrence predictions in Venezuelan Psittacidae

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Iniciativa para el Mapeo de la Biodiversidad Neotropical

Monitoring Psittacidae in Venezuela: Distribution data and occupancy models for 2010



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1 Introduction

Venezuela has a high diversity of psittacids (Aves, Psittaciformes) but many species are threatened by increasing rate of land transformation (Oliveira-Miranda *et al.*, 2010) and illegal wildlife trade (Sánchez-Mercado *et al.*, 2017). The UICN reports declining regional trend for 34 of the 50 Psittacidae species occurring in the country, and six species are already under some threat category in the Venezuelan Red Data Book (Rodríguez & Rojas-Suárez, 2008).

Here we provide an assessment of current distribution of Psittacid species in Venezuela using data from a national bird monitoring program carried out in 2010 as a component of the Neotropical Biodiversity Mapping Initiative (NeoMaps Ferrer-Paris *et al.*, 2013a; Rodríguez & Sharpe, 2002; Rodríguez *et al.*, 2012).

2 Methods

2.1 Distribution and monitoring data

Information on bird species present in Venezuela is taken from (Hilty, 2003). Range maps for all these species were obtained from BirdLife International (BirdLife International & Nature-Serve, 2014). Distribution records were retrieved from the Global Biodiversity Information Facility (GBIF.org, 2016).

NeoMaps bird survey was completed between March and April 2010 (Rodríguez *et al.*, 2012). Sampling universe consisted in 170 half-degree cells defined in the Venezuelan Biodiversity Grid, which cover over half of the country, but does not include the southern forest regions. Twenty seven cells were selected using a stratified sampling design based on environmental and biogeographical variables. Number in the following figure refer to NeoMaps transect codes (NM01 ... NM93).



Standardized field sampling protocols for birds was implemented along a 40 km roadside transect within each cell. Two surveys were performed during two consecutive days in each transect: on the first day, 3-min point counts were performed at 50 stops, 800m apart. On the second day, cumulative species lists were recorded at a selection of 10 stops sampled for 9 min each, divided into three consecutive 3-min periods. Total sampling effort was 108 hours

of bird surveys (Rodríguez et al., 2011; Rodríguez et al., 2012)

For this analysis we built detection histories for each psittacids species recorded by NeoMapas. We considered each stop as a "site" (*i*; 1351 sites, 50 stops across 27 transects), and each timed survey period of 3 min as a "observation" (*j*), with duration d = 3min. For the first day survey, detections were recorded as "1" and lack of detections as "0". For the cumulative list of the second day the detection history was filled with "0" until the first detection, and with null values (*N*) afterward. Thus valid detection histories for the second day are 1*NN*, 01*N*, 001 and 000, or *NNN* if the site was not visited on the second day (Ferrer-Paris *et al.*, 2013b). Time of the day was used as an observation covariate. Sites covariates were extracted from the spatial location of each site.

2.2 Sites covariates

In order to get representative data on climatic and vegetation condition at the time of the survey, we matched the location and date of each observations with time-series of environmental variables derived from the Moderate Resolution Radio Spectrometer (MODIS) sensors in Terra-Satellites and and queried using the global MODIS Subsetting Tool (Land Processes Distributed Active Archive Center), and the Climate Hazards Group InfraRed Precipitation with Station data archive (CHIRPS version 2.0). We calculated the representative value of the variable for the year prior to the sampling time (approx. march 2009 - march 2010). We used the following variables:

2.2.1 Total annual precipitation

Total precipitation of the year prior to NM sampling according to Chirps v 2.0(Funk et al., 2015, pre01)



2.2.2 Potential Evapotranspiration

Total Potential Evapotranspiration of the year prior to NM (Mu et al., 2011, pet01).



2.2.3 Land Surface Temperature

Mean LST of the year prior to NM (Wan *et al.*, 2004, dT01).



2.2.4 Enhanced Vegetation Index

Mean EVI of the year prior to NM sampling (Huete et al., 2002, evi01), :



2.3 Occupancy models

We used a single-season occupancy model based on zero-inflated binomial models (MacKenzie *et al.*, 2006) to estimate a probability of occurrence for species detected in the surveys (Ψ) . The occupancy state (z_i) of site *i* was modeled as z_i Bernoulli (Ψ_i) , while the observation process was modeled as $y_{ij}|z_i$ Bernoulli $(z_i * p_{ij})$ in which $p_i j$ represented site and occasion specific detection probability. Covariates of Ψ_i (site covariates) and p_{ij} (observation covariates) were modeled using the logit link.

We fitted eight models representing different combinations of covariates for probability of detection and probability of occurrence:

$ \begin{array}{lll} \mbox{nulo} & p \sim 1 & \Psi \sim 1 \\ p(h)Psi(.) & p \sim h & \Psi \sim 1 \\ p(.)Psi(V) & p \sim 1 & \Psi \sim \mbox{evi}01 + \mbox{evi}01^2 \\ p(h)Psi(V) & p \sim h & \Psi \sim \mbox{evi}01 + \mbox{evi}01^2 \\ p(.)Psi(C) & p \sim 1 & \Psi \sim \mbox{dT01} + \mbox{dT01}^2 + \mbox{pre01} + \mbox{pre01}^2 + \mbox{pet01} + \mbox{pet01}^2 \\ p(h)Psi(C) & p \sim h & \Psi \sim \mbox{dT01} + \mbox{dT01}^2 + \mbox{pre01} + \mbox{pre01}^2 + \mbox{pet01} + \mbox{pet01}^2 \\ p(.)Psi(VC) & p \sim h & \Psi \sim \mbox{evi}01 + \mbox{evi}01^2 + \mbox{dT01} + \mbox{dT01}^2 + \mbox{pre01} + \mbox{pre01}^2 + \mbox{pet01} + \mbox{pet01}^2 \\ p(h)Psi(VC) & p \sim h & \Psi \sim \mbox{evi}01 + \mbox{evi}01^2 + \mbox{dT01} + \mbox{dT01}^2 + \mbox{pre01} + \mbox{pre01}^2 + \mbox{pet01} + \mbox{pet01}^2 \\ p(h)Psi(VC) & p \sim h & \Psi \sim \mbox{evi}01 + \mbox{evi}01^2 + \mbox{dT01} + \mbox{dT01}^2 + \mbox{pre01} + \mbox{pre01}^2 + \mbox{pet01} + \mbox{pet01}^2 \\ p(h)Psi(VC) & p \sim h & \Psi \sim \mbox{evi}01 + \mbox{evi}01^2 + \mbox{dT01} + \mbox{dT01}^2 + \mbox{pre01} + \mbox{pre01}^2 + \mbox{pet01} + \mbox{pet01}^2 \\ p(h)Psi(VC) & p \sim h & \Psi \sim \mbox{evi}01 + \mbox{evi}01^2 + \mbox{dT01} + \mbox{dT01}^2 + \mbox{pre01} + \mbox{pet01}^2 + \mbox{pet01} + \mbox{pet01}^2 \\ p(h)Psi(VC) & p \sim h & \Psi \sim \mbox{evi}01 + \mbox{evi}01^2 + \mbox{dT01} + \mbox{dT01}^2 + \mbox{pre01} + \mbox{pet01}^2 + \mbox{pet01}^2 \\ p(h)Psi(VC) & p \sim h & \Psi \sim \mbox{evi}01 + \mbox{evi}01^2 + \mbox{dT01} + \mbox{dT01}^2 + \mbox{pre01} + \mbox{pet01}^2 \\ p(h)Psi(VC) & p \sim h & \Psi \sim \mbox{evi}01 + \mbox{evi}01^2 + \mbox{dT01} + \mbox{dT01}^2 + \mbox{pre01} + \mbox{pet01}^2 \\ p(h)Psi(VC) & p \sim h & \Psi \sim \mbox{evi}01 + \mbox{evi}01^2 + \mbox{dT01} + \mbox{dT01}^2 + \mbox{pre01} + \mbox{pet01}^2 \\ p(h)Psi(VC) & p \sim h & \Psi \sim \mbox{evi}01 + \mbox{evi}01^2 + \mbox{dT01} + \m$	Name	Detection	Presence
$\begin{array}{lll} p(h)Psi(.) & p \sim h & \Psi \sim 1 \\ p(.)Psi(V) & p \sim 1 & \Psi \sim \text{evi01} + \text{evi01}^2 \\ p(h)Psi(V) & p \sim h & \Psi \sim \text{evi01} + \text{evi01}^2 \\ p(.)Psi(C) & p \sim h & \Psi \sim \text{dT01} + \text{dT01}^2 + \text{pre01} + \text{pre01}^2 + \text{pet01} + \text{pet01}^2 \\ p(h)Psi(C) & p \sim h & \Psi \sim \text{dT01} + \text{dT01}^2 + \text{pre01} + \text{pre01}^2 + \text{pet01} + \text{pet01}^2 \\ p(.)Psi(VC) & p \sim h & \Psi \sim \text{evi01} + \text{evi01}^2 + \text{dT01} + \text{dT01}^2 + \text{pre01} + \text{pre01}^2 + \text{pet01} + \text{pet01}^2 \\ p(h)Psi(VC) & p \sim h & \Psi \sim \text{evi01} + \text{evi01}^2 + \text{dT01} + \text{dT01}^2 + \text{pre01} + \text{pre01}^2 + \text{pet01} + \text{pet01}^2 \\ \end{array}$	nulo	$p \sim 1$	$\Psi \sim 1$
$\begin{array}{lll} p(.)Psi(V) & p \sim 1 & \Psi \sim \text{evi01} + \text{evi01}^2 \\ p(h)Psi(V) & p \sim h & \Psi \sim \text{evi01} + \text{evi01}^2 \\ p(.)Psi(C) & p \sim 1 & \Psi \sim \text{dT01} + \text{dT01}^2 + \text{pre01} + \text{pre01}^2 + \text{pet01} + \text{pet01}^2 \\ p(h)Psi(C) & p \sim h & \Psi \sim \text{dT01} + \text{dT01}^2 + \text{pre01} + \text{pre01}^2 + \text{pet01} + \text{pet01}^2 \\ p(.)Psi(VC) & p \sim h & \Psi \sim \text{evi01} + \text{evi01}^2 + \text{dT01} + \text{dT01}^2 + \text{pre01} + \text{pre01}^2 + \text{pet01} + \text{pet01}^2 \\ p(h)Psi(VC) & p \sim h & \Psi \sim \text{evi01} + \text{evi01}^2 + \text{dT01} + \text{dT01}^2 + \text{pre01} + \text{pre01}^2 + \text{pet01} + \text{pet01}^2 \\ \end{array}$	p(h)Psi(.)	$p \sim h$	$\Psi \sim 1$
$\begin{array}{lll} p(h)Psi(V) & p \sim h & \Psi \sim \text{evi01} + \text{evi01}^2 \\ p(.)Psi(C) & p \sim 1 & \Psi \sim \text{dT01} + \text{dT01}^2 + \text{pre01} + \text{pre01}^2 + \text{pet01} + \text{pet01}^2 \\ p(h)Psi(C) & p \sim h & \Psi \sim \text{dT01} + \text{dT01}^2 + \text{pre01} + \text{pre01}^2 + \text{pet01} + \text{pet01}^2 \\ p(.)Psi(VC) & p \sim h & \Psi \sim \text{evi01} + \text{evi01}^2 + \text{dT01} + \text{dT01}^2 + \text{pre01} + \text{pre01}^2 + \text{pet01} + \text{pet01}^2 \\ p(h)Psi(VC) & p \sim h & \Psi \sim \text{evi01} + \text{evi01}^2 + \text{dT01} + \text{dT01}^2 + \text{pre01} + \text{pre01}^2 + \text{pet01} + \text{pet01}^2 \end{array}$	p(.)Psi(V)	$p \sim 1$	$\Psi \sim \mathrm{evi}01 + \mathrm{evi}01^2$
$\begin{array}{ll} p(.)Psi(C) & p \sim 1 & \Psi \sim \mathrm{dT01} + \mathrm{dT01}^2 + \mathrm{pre01} + \mathrm{pre01}^2 + \mathrm{pet01} + \mathrm{pet01}^2 \\ p(h)Psi(C) & p \sim h & \Psi \sim \mathrm{dT01} + \mathrm{dT01}^2 + \mathrm{pre01} + \mathrm{pre01}^2 + \mathrm{pet01} + \mathrm{pet01}^2 \\ p(.)Psi(VC) & p \sim h & \Psi \sim \mathrm{evi01} + \mathrm{evi01}^2 + \mathrm{dT01} + \mathrm{dT01}^2 + \mathrm{pre01} + \mathrm{pre01}^2 + \mathrm{pet01} + \mathrm{pet01}^2 \\ p(h)Psi(VC) & p \sim h & \Psi \sim \mathrm{evi01} + \mathrm{evi01}^2 + \mathrm{dT01} + \mathrm{dT01}^2 + \mathrm{pre01} + \mathrm{pre01}^2 + \mathrm{pet01} + \mathrm{pet01}^2 \end{array}$	p(h)Psi(V)	$p \sim h$	$\Psi \sim \text{evi}01 + \text{evi}01^2$
$ \begin{array}{ll} p(h)Psi(C) & p \sim h & \Psi \sim dT01 + dT01^2 + pre01 + pre01^2 + pet01 + pet01^2 \\ p(.)Psi(VC) & p \sim h & \Psi \sim evi01 + evi01^2 + dT01 + dT01^2 + pre01 + pre01^2 + pet01 + pet01^2 \\ p(h)Psi(VC) & p \sim h & \Psi \sim evi01 + evi01^2 + dT01 + dT01^2 + pre01 + pre01^2 + pet01 + pet01^2 \\ \end{array} $	p(.)Psi(C)	$p \sim 1$	$\Psi \sim dT01 + dT01^2 + pre01 + pre01^2 + pet01 + pet01^2$
$ \begin{array}{ll} p(.)Psi(VC) & p \sim h & \Psi \sim \text{evi}01 + \text{evi}01^2 + \text{dT}01 + \text{dT}01^2 + \text{pre}01 + \text{pre}01^2 + \text{pet}01 + \text{pet}01^2 \\ p(h)Psi(VC) & p \sim h & \Psi \sim \text{evi}01 + \text{evi}01^2 + \text{dT}01 + \text{dT}01^2 + \text{pre}01 + \text{pre}01^2 + \text{pet}01 + \text{pet}01^2 \\ \end{array} $	p(h)Psi(C)	$p \sim h$	$\Psi \sim dT01 + dT01^2 + pre01 + pre01^2 + pet01 + pet01^2$
$p(h)Psi(VC)$ $p \sim h$ $\Psi \sim evi01 + evi01^2 + dT01 + dT01^2 + pre01 + pre01^2 + pet01 + pet01^2$	p(.)Psi(VC)	$p \sim h$	$\Psi \sim evi01 + evi01^2 + dT01 + dT01^2 + pre01 + pre01^2 + pet01 + pet01^2$
	p(h)Psi(VC)	$p \sim h$	$\Psi \sim evi01 + evi01^2 + dT01 + dT01^2 + pre01 + pre01^2 + pet01 + pet01^2$

We evaluated the individual performance of each model using the corrected Akaike Information Criterion (AICc). Then, we used the model with the best performance for each species to explain the lack of detections across the survey sites. For the sites without detections, we calculated the conditional probability of occurrence given that the species was not detected using empirical bayes methods (MacKenzie *et al.*, 2006). This probability (Ψ_{condl}) considers two components: whether sampling effort was enough to detect the species at least once conditional on its presence $(p* = 1 - \prod(1-p))$, and the unconditional probability of occurrence given the values of the site covariates $(\hat{\Psi})$.

We used the unmarked, raster, and AICcmodavg packages of R to fit the models (Fiske & Chandler, 2011; Hijmans, 2017; Mazerolle, 2017; R Core Team, 2017)

3 Species accounts

3.1 Genus Amazona

The genus Amazona is represented in Venezuela by eight species. The following table show species names and acronyms used in this text, with the number of distribution for each species according to the Global Biodiversity Information Facility (GBIF), and the number of records from the 2008-2012 period (GBIF.2010), and the number of detection in NeoMaps first day sampling (NM.M1), and the additional detections in the three sampling periods of the second day (NM.L1,NM.L2,NM.L3).

		aspp	GBIF	GBIF.2010	NM.M1	NM.L1	NM.L2	NM.L3
Amazona	amazonica	Amaz_amaz	547	120	75	13	6	7
Amazona	autumnalis	Amaz_autu	12	3	NA	NA	NA	NA
Amazona	barbadensis	Amaz_barb	40	12	3	2	0	1
Amazona	bodini	Amaz_bodi	10	0	NA	NA	NA	NA
Amazona	dufresniana	Amaz_dufr	36	7	NA	NA	NA	NA
Amazona	farinosa	Amaz_fari	212	46	12	3	1	1
Amazona	mercenarius	Amaz_merc	13	1	NA	NA	NA	NA
Amazona	ochrocephala	Amaz_ochr	719	264	108	33	4	5

Only four species were detected during NeoMaps surveys in 2010, but the four undetected species also had very few GBIF records in this time period. Please refer to Ferrer-Paris *et al.* (2013b,c) for alternative model parametrization and detailed data for this genus.

3.1.1 Amazona amazonica

Amazona amazonica is a widespread species. The following maps shows the expected distribution in Venezuela and surrounding countries according to BirdLife polygon maps (pink polygons) and curated distribution records from GBIF for the time period of 2008-2012. Blue dots represents detection of the species and red dots represent detection of other species of Psittacidae, but lack of detection of the target species.



Some GBIF records are found outside the polygon of expected distribution. This can be an indication of incomplete representation of the known distribution in BirdLife maps, or taxonomic uncertainty in identification of GBIF records.

For this species we could fit several alternative occupancy models based on the large number of detections during the 2010 field survey data in Venezuela. We compare these models in the following table, were mod is the model description, n is the sample size (number of localities within its expected distribution or with evidence of presence), dtt is the number of detections, AICc is the corrected Akaike Information Criteria, Delta.AICc is the difference in AICc to the model with lowest AICc, AICw are the Akaike weights and LL is the log likelihood.

	spp	mod	n	dtt	AICc	Delta.AICc	AICw	LL
1	Amaz_amaz	p(h)Psi(VC)	900	97	553.7420	0.00	0.999	-265.72
2	Amaz_amaz	p(h)Psi(C)	900	97	567.4946	13.75	0.001	-274.65
3	Amaz_amaz	p(.)Psi(VC)	900	97	673.6545	119.91	0.000	-326.70
4	Amaz_amaz	p(.)Psi(C)	900	97	696.9515	143.21	0.000	-340.39
5	Amaz_amaz	p(.)Psi(V)	900	97	712.8495	159.11	0.000	-352.40
6	Amaz_amaz	nulo	900	97	726.6031	172.86	0.000	-361.29

For this species the p(h)Psi(VC) model had the highest support according to the AIC weights. This model corresponds to a time-dependent probability of detection and a effect of vegetation and climatic conditions on the probability of presence. The following figure shows a weighted estimate of probability of detection per hour of the day for the two models with AICw > 0.



The following map shows the predicted (unconditional) probability of presence for the whole country based on the model with highest support and the values of the vegetation and climatic covariates. Darker colors indicate higher probabilities, dots represent NeoMaps sampling localities, blue dots indicate detections and red dots indicate lack of detections.



3.1.2 Amazona autumnalis

This species was expected in 50 sampling points from 'Rosario de Perijá', Zulia state (NeoMaps route NM05). However this species was not detected during the field work in 2010.



Known GBIF records are found outside the polygon of expected distribution. This can be an indication of incomplete representation of the known distribution in BirdLife maps, or taxonomic uncertainty in identification of GBIF records.

3.1.3 Amazona barbadensis

Amazona barbadensis is a species of restricted distribution in Venezuela and the Caribbean.



During NeoMaps field surveys in 2010, *Amazona barbadensis* was detected in few localities, and due to small sample size, only two models could be fitted, both have similar support according to AIC weights.

	spp	mod	n	dtt	AICc	Delta.AICc	AICw	LL
7	Amaz_barb	p(h)Psi(.)	150	5	55.07084	0.00	0.571	-24.45
8	Amaz_barb	nulo	150	5	55.64423	0.57	0.429	-25.78

The following figure shows a weighted estimate of probability of detection per hour of the day for the two models with AICw > 0.



Unconditional probability of presence is relatively low (around 0, 2). For the three NeoMaps routes were the species was expected, presence was confirmed in very few localities (localities of detection), and true absence can be suspected in around 18% of the sampling localities were the sampling effort was enough to achieve low values of the posterior or conditional probability of presence ($\Psi_{post} < 0, 12$).



3.1.4 Amazona bodini

Amazona bodini is a species of restricted distribution in Venezuela. It was not detected by NeoMaps field work, and is not represented in GBIF record from the 2008-2012 period.

Amazona bodini was expected in 102 sampling points from 'Isla de Guara', Monagas-Delta

state (NeoMaps route NM36); 'Caicara del Orinoco', Bolívar state (NeoMaps route NM75). However this species was not detected during the field work in 2010.



3.1.5 Amazona dufresniana

Amazona dufresniana is a species of restricted distribution in Venezuela. It was not detected by NeoMaps field work, and has few GBIF record from the 2008-2012 period.

Amazona dufresniana was expected in 109 sampling points from 'Anacoco', Bolívar state (NeoMaps route NM93); 'La Escalera', Bolívar state (NeoMaps route NM90); 'Kavanayén', Bolívar state (NeoMaps route NM21). However this species was not detected during the field work in 2010.



3.1.6 Amazona farinosa

Amazona farinosa has a non-continuous distribution in Venezuela and was expected in 482 sampling points from 'Rosario de Perijá', Zulia state (NeoMaps route NM05); 'Sur del Lago', Zulia-Mérida state (NeoMaps route NM07); 'Guri', Bolívar state (NeoMaps route NM24); 'El Manteco', Bolívar state (NeoMaps route NM22); 'Caicara del Orinoco', Bolívar state (NeoMaps route NM75); 'Gavilán', Amazonas state (NeoMaps route NM34); 'Anacoco', Bolívar state (NeoMaps route NM93); 'La Escalera', Bolívar state (NeoMaps route NM90); 'Kavanayén', Bolívar state (NeoMaps route NM21); 'Paraytepuy', Bolívar state (NeoMaps route NM23).



Available data allowed to fit models with constant probability of detection. The model with highest support according to AIC weights includes climatic and vegetion condition as covariates.

	spp	mod	n	dtt	AICc	Delta.AICc	AICw	LL
9	Amaz_fari	p(.)Psi(VC)	550	17	158.5461	0.00	0.940	-69.07
10	Amaz_fari	p(.)Psi(V)	550	17	164.0841	5.54	0.059	-78.01
11	Amaz_fari	p(.)Psi(C)	550	17	171.3484	12.80	0.002	-77.54

According to the model with the best support the unconditional probability of presence varied along the NeoMaps routes depending on the value of the covariates. The figure show four examples of 40 km routes, ordered from beginning to end, two with detections (NM05 and NM18) and two without detections (NM75 and NM93). The lines show the predicted unconditional probability of presence (solid line: best estimate, dotted lines 95% confidence interval), and the dots represent the conditional probability given the observed detection history. Blue dots represent the localities of known occurrences, red dots are localities with very low posterior probabilities of presence ($\Psi_{post} < 0.125$), grey dots are localities with intermediate values.



The predicted (unconditional) probability of presence for the whole country does reflect the expected distribution, but predicts high probability of presence in areas of Falcón and Lara state where the species is absent, probably due to biogeographic constrains.



3.1.7 Amazona mercenarius

Amazona mercenarius was expected in 103 sampling points from 'Jají', Mérida state (NeoMaps route NM26); 'Piñango', Mérida state (NeoMaps route NM57). However this species was not detected during the field work in 2010.



3.1.8 Amazona ochrocephala

Amazona ochrocephala has a widespread distribution in Venezuela and was detected in several NeoMaps transects.



Available data allowed to fit several models with constant probability of detection. The model with highest support according to AIC weights includes climatic and vegetion condition as covariates.

	spp	mod	n	dtt	AICc	Delta.AICc	AICw	LL
12	$Amaz_ochr$	p(.)Psi(VC)	1250	143	888.5061	0.00	1	-434.16
13	Amaz_ochr	p(.)Psi(C)	1250	143	916.9785	28.47	0	-450.43
14	Amaz_ochr	p(.)Psi(V)	1250	143	987.4805	98.97	0	-489.72
15	Amaz_ochr	nulo	1250	143	1042.1294	153.62	0	-519.06

The predicted (unconditional) probability of presence for the whole country does reflect the expected distribution.



3.2 Genus Ara

The genus Ara is represented in Venezuela by five species.

		aspp	GBIF	GBIF.2010	NM.M1	NM.L1	NM.L2	NM.L3
Ara	ararauna	Ara_arar	140	56	NA	NA	NA	NA
Ara	chloropterus	Ara_chlo	271	52	4	4	4	1
Ara	macao	Ara_maca	290	117	6	1	1	1
Ara	militaris	Ara_mili	80	18	6	0	2	0
Ara	severus	Ara_seve	457	217	48	11	3	2

Four species were detected during NeoMaps surveys in 2010, the only exception was (Ara ararauna).

3.2.1 Ara ararauna

Ara ararauna was expected in 112 sampling points from 'Isla de Guara', Monagas-Delta state (NeoMaps route NM36); 'Gavilán', Amazonas state (NeoMaps route NM34); 'Paraytepuy', Bolívar state (NeoMaps route NM23). However this species was not detected during the field work in 2010.



3.2.2 Ara chloropterus

Ara chloropterus has a widespread distribution and was expected in 1045 sampling points from NeoMaps, but has few actual detections.



Available data allowed to fit models with constant probability of detection and variable probability of presence.

	spp	mod	n	dtt	AICc	Delta.AICc	AICw	LL
16	Ara_chlo	p(h)Psi(.)	1050	13	149.4942	0.00	0.964	-71.74
17	Ara_chlo	nulo	1050	13	156.0788	6.58	0.036	-76.03

However, the model has a poor fit with a combination of high probability of occurrence and low probability of detection and too wide standard errors for the estimated parameters:

Call: occu(formula = ~hora ~ 1, data = UMF[os,])

```
Occupancy (logit-scale):
 Estimate
            SE
                   z P(|z|)
     3.99 25.7 0.155
                       0.877
Detection (logit-scale):
            Estimate
                       SE
                               z P(|z|)
(Intercept)
              -0.291 1.74 -0.168
                                  0.8666
             -13.706 5.34 -2.568 0.0102
hora
AIC: 149.4713
Number of sites: 1050
optim convergence code: 0
optim iterations: 63
Bootstrap iterations: 0
```

3.2.3 Ara macao

Ara macao has a widespread distribution and was expected in 607 sampling points from NeoMaps, but has few actual detections.



Available data allowed to fit several alternative models including covariates in probability of presence.

	\mathtt{spp}	mod	n	dtt	AICc	Delta.AICc	AICw	LL
18	Ara_maca	p(.)Psi(C)	600	8	79.22826	0.00	0.692	-31.49
19	Ara_maca	p(h)Psi(C)	600	8	80.85336	1.63	0.307	-31.27
20	Ara_maca	p(.)Psi(V)	600	8	93.83741	14.61	0.000	-42.89
21	Ara_maca	nulo	600	8	100.99736	21.77	0.000	-48.49
22	Ara_maca	p(h)Psi(.)	600	8	102.79301	23.56	0.000	-48.38

The best model includes constant detection probability and effect on climatic covariates, but parameter values suggest a poor fit and high uncertainty.

```
Call:
occu(formula = ~1 ~ pet01 + I(pet01^2) + dT01 + I(dT01^2) + pre01 +
    I(pre01^2), data = UMF[os, ])
Occupancy (logit-scale):
            Estimate
                        SE
                                  z P(|z|)
(Intercept)
             -4.7783
                      2.11 -2.2634
                                    0.0236
              2.4933
                      2.23
                            1.1202
pet01
                                    0.2626
I(pet01^2)
              0.0817
                      1.13
                            0.0722
                                    0.9424
dT01
                      4.41
                            1.3627
              6.0089
                                     0.1730
I(dT01^2)
                      2.85 -0.8835
             -2.5151
                                    0.3770
pre01
              0.8769 11.11 0.0789
                                    0.9371
            -16.0437 24.46 -0.6560
I(pre01^2)
                                    0.5118
Detection (logit-scale):
 Estimate
             SE
                    z P(|z|)
    -1.04 0.775 -1.35
                        0.179
AIC: 78.9846
Number of sites: 600
optim convergence code: 0
optim iterations: 74
Bootstrap iterations: 0
```

Unconditional probability of presence is relatively low in most of the sampling area, and presence was confirmed in very few localities (localities of detection).



Histogram of psi.cond

3.2.4 Ara militaris

Ara militaris is found in the Andean and Coastal mountain ranges of Venezuela, and was expected in 35 sampling points from 'Rosario de Perijá', Zulia state (NeoMaps route NM05); 'Colonia Tovar', Aragua-Vargas state (NeoMaps route NM80).



Available data allowed to fit some alternative models with similar AICc values.

	spp	mod	n	dtt	AICc	Delta.AICc	AICw	LL
23	Ara_mili	p(h)Psi(V)	150	8	71.80744	0.00	0.365	-30.70
24	Ara_mili	p(.)Psi(V)	150	8	72.38796	0.58	0.273	-32.06
25	Ara_mili	p(h)Psi(.)	150	8	72.60602	0.80	0.245	-33.22
26	Ara_mili	nulo	150	8	74.09608	2.29	0.116	-35.01

The best model includes an effect of time of day on the detection probability.



Parameter estimates have high values and high uncertainty.

```
Call:
occu(formula = ~hora ~ evi01 + I(evi01^2), data = UMF[os, ])
Occupancy (logit-scale):
            Estimate
                       SE
                                z P(|z|)
(Intercept)
              -3.388 2.89 -1.173
                                    0.241
evi01
               0.996 5.48
                            0.182
                                    0.856
I(evi01^2)
               0.761 2.49
                           0.305
                                    0.760
Detection (logit-scale):
            Estimate
                                z P(|z|)
                       SE
(Intercept)
                1.95 2.66
                           0.732
                                    0.464
              -10.59 6.90 -1.534
hora
                                    0.125
AIC: 71.39078
Number of sites: 150
optim convergence code: 0
optim iterations: 41
Bootstrap iterations: 0
```

Unconditional probability of presence is relatively low in most of the sampling area, and presence was confirmed in very few localities (localities of detection).

Histogram of psi.cond



3.2.5 Ara severus

Ara severus has a widespread but non continuos distribution and was expected in 642 sampling points from NeoMaps.



Available data allowed to fit models with constant probability of detection and probability of presence.

	spp	mod	n	dtt	AICc	Delta.AICc	AICw		LL
27	Ara_seve	p(.)Psi(V)	700	61	450.5242	0.00	1	-221	. 23
28	Ara_seve	nulo	700	61	482.0490	31.52	0	-239	. 02

Parameter estimates have high values and high uncertainty.

```
Call:
occu(formula = ~1 ~ evi01 + I(evi01^2), data = UMF[os, ])
Occupancy (logit-scale):
            Estimate
                        SE
                                z P(|z|)
                                    0.497
                4.22
                      6.21 0.679
(Intercept)
evi01
               11.59 10.41 1.114
                                    0.265
I(evi01^2)
               -7.84 8.13 -0.964
                                    0.335
Detection (logit-scale):
 Estimate
             SE
                    z P(>|z|)
    -2.37 0.145 -16.3 1.25e-59
AIC: 450.4667
Number of sites: 700
optim convergence code: 0
optim iterations: 81
Bootstrap iterations: 0
```

The combination of parameter values results in extreme predictions of conditional probability of presence within the sampling area.





3.3 Genus Aratinga

Evidence for the presence of the genus Aratinga in Venezuela is scarce. One species is suspected to be present, but reliable records are missing.

		aspp	GBIF	GBIF.2010	NM.M1	NM.L1	NM.L2	NM.L3
Aratinga	solstitialis	Arat_sols	1	0	NA	NA	NA	NA

3.3.1 Aratinga solstitialis

The distribution of *Aratinga solstitialis* probably includes part of the Guayana Esequiba, but taxonomic and distribution information for this taxon needs further review (Silveira *et al.*, 2005).



3.4 Genus Bolborhynchus, species B. lineola

The genus Bolborhynchus is represented in Venezuela by a single species.

			aspp	GBIF	GBIF.201	O NM.M1	NM.L1	NM.L2	NM.L3
Bolborhynchus	lineola	Bolb	line	30	2	2 NA	NA	NA	NA

Bolborhynchus lineola is a species of restricted distribution in Venezuela. Bolborhynchus lineola was expected in 96 sampling points from 'Sur del Lago', Zulia-Mérida state (NeoMaps route NM07); 'Jají', Mérida state (NeoMaps route NM26); 'San Joaquín de Navay', Táchira state (NeoMaps route NM18); 'Colonia Tovar', Aragua-Vargas state (NeoMaps route NM80). It was not detected by NeoMaps field work, and only has few GBIF record from the 2008-2012 period.



3.5 Genus Brotogeris

The genus Brotogeris is represented in Venezuela by three species, but only two were detected in NeoMaps surveys of 2010.

		aspp	GBIF	GBIF.2010	NM.M1	NM.L1	NM.L2	NM.L3
Brotogeris	chrysoptera	Brot_chry	51	14	0	0	2	0
Brotogeris	cyanoptera	Brot_cyan	27	15	NA	NA	NA	NA
Brotogeris	jugularis	Brot_jugu	412	218	16	2	3	0

3.5.1 Brotogeris chrysoptera

Brotogeris chrysoptera is only found in eastern Venezuela, and was expected in 454 sampling points from 'Paria', Sucre state (NeoMaps route NM45); 'San Tomé', Anzoátegui-Monagas state (NeoMaps route NM14); 'Isla de Guara', Monagas-Delta state (NeoMaps route NM36); 'Guri', Bolívar state (NeoMaps route NM24); 'El Manteco', Bolívar state (NeoMaps route NM22); 'Anacoco', Bolívar state (NeoMaps route NM93); 'La Escalera', Bolívar state (NeoMaps route NM90); 'Kavanayén', Bolívar state (NeoMaps route NM21); 'Paraytepuy', Bolívar state (NeoMaps route NM23).



It was detected twice during the field work in 2010. Due to the scarce data available, only two models could be fitted. The null model had best support according to AIC weights, but a poor fit with a combination of high probability of occurrence and low probability of detection and too wide standard errors for the estimated parameters:

```
LL
                    mod
                          n dtt
                                     AICc Delta.AICc AICw
         spp
40 Brot_chry
                   nulo 450
                               2 31.53931
                                                0.00 0.734 -13.76
41 Brot_chry p(h)Psi(.) 450
                              2 33.56471
                                                2.03 0.266 -13.76
Call:
occu(formula = ~1 ~ 1, data = UMF[os, ])
Occupancy (logit-scale):
 Estimate SE
                  z P(|z|)
     2.26 26 0.0869
                      0.931
Detection (logit-scale):
 Estimate
            SE
                   z P(|z|)
    -5.78 2.56 -2.26
                       0.024
AIC: 31.51246
Number of sites: 450
optim convergence code: 0
optim iterations: 82
Bootstrap iterations: 0
```

3.5.2 Brotogeris cyanoptera

Brotogeris cyanoptera is only found in souther Venezuela in Amazonas state, and was expected in 50 sampling points from 'Gavilán', Amazonas state (NeoMaps route NM34). It was not detected during the field work in 2010.



3.5.3 Brotogeris jugularis

Brotogeris jugularis is only found in western Venezuela, and was expected in 557 sampling points from 'Rosario de Perijá', Zulia state (NeoMaps route NM05); 'Quebrada Arriba', Lara-Falcón state (NeoMaps route NM03); 'Sur del Lago', Zulia-Mérida state (NeoMaps route NM07); 'Jají', Mérida state (NeoMaps route NM26); 'Piñango', Mérida state (NeoMaps route NM57); 'Capitanejo', Barinas state (NeoMaps route NM19); 'Otopún', Barinas state (NeoMaps route NM16); 'San Joaquín de Navay', Táchira state (NeoMaps route NM18); 'Paraguaná', Falcón state (NeoMaps route NM02); 'Río Tocuyo', Lara state (NeoMaps route NM01); 'Colonia Tovar', Aragua-Vargas state (NeoMaps route NM80); 'Corralito', Cojedes state (NeoMaps route NM96); 'Altagracia de Orituco', Aragua state (NeoMaps route NM08).



Available data allowed to fit several alternative models, and models with either climatic or vegetation covariates received considerable support according to AIC weights.

	spp	mod	n	dtt	AICc	Delta.AICc	AICw	LL
42	Brot_jugu	p(h)Psi(C)	650	20	174.6283	0.00	0.551	-78.17
43	Brot_jugu	p(.)Psi(C)	650	20	175.4298	0.80	0.369	-79.60
					05			

44	Brot_jugu	p(.)Psi(V)	650	20	179.1705	4.54	0.057	-85.55
45	Brot_jugu	p(h)Psi(V)	650	20	180.9018	6.27	0.024	-85.40
46	Brot_jugu	nulo	650	20	199.9915	25.36	0.000	-97.99
47	Brot_jugu	p(h)Psi(.)	650	20	200.7790	26.15	0.000	-97.37

The best model includes an effect of time of day on the detection probability.



Hour of the day

```
Call:
occu(formula = ~hora ~ pet01 + I(pet01^2) + dT01 + I(dT01^2) +
    pre01 + I(pre01^2), data = UMF[os, ])
Occupancy (logit-scale):
            Estimate
                        SE
                                 z P(>|z|)
(Intercept)
               -1.11 0.573 -1.934
                                   0.0531
                0.35 0.755 0.463
pet01
                                   0.6433
I(pet01^2)
               -1.60 0.976 -1.637
                                    0.1016
dT01
               -1.69 0.660 -2.564
                                   0.0103
I(dT01^2)
               -2.23 0.903 -2.464
                                   0.0137
pre01
                1.10 0.555 1.988
                                   0.0468
I(pre01^2)
               -0.37 0.203 -1.821
                                   0.0685
Detection (logit-scale):
            Estimate
                       SE
                               z P(|z|)
(Intercept)
                3.51 2.42
                           1.45
                                   0.147
hora
              -10.17 6.17 -1.65
                                   0.099
AIC: 174.3471
```

Number of sites: 650

```
optim convergence code: 0
optim iterations: 33
Bootstrap iterations: 0
```

The following map shows the predicted (unconditional) probability of presence for the whole country based on the model with highest support and the values of the climatic covariates.



3.6 Genus Deroptyus, species *D. accipitrinus*

The genus Deroptyus is represented in Venezuela by a single species.

		aspp	GBIF	GBIF.2010	NM.M1	NM.L1	NM.L2	NM.L3
Deroptyus	accipitrinus	Dero_acci	49	11	NA	NA	NA	NA

Deroptyus accipitrinus is a species of restricted to South Venezuela, and was expected in 308 sampling points from 'Guri', Bolívar state (NeoMaps route NM24); 'El Manteco', Bolívar state (NeoMaps route NM22); 'Gavilán', Amazonas state (NeoMaps route NM34); 'Anacoco', Bolívar state (NeoMaps route NM93); 'La Escalera', Bolívar state (NeoMaps route NM90); 'Kavanayén', Bolívar state (NeoMaps route NM21); 'Paraytepuy', Bolívar state (NeoMaps route NM23). However this species was not detected during the field work in 2010, and only has few GBIF record from the 2008-2012 period.



3.7 Genus Diopsittaca, species D. nobilis

This genus is represented in Venezuela by a single species.

aspp	GBIF	GBIF.2010	NM.M1	NM.L1	NM.L2	NM.L3
Diopsittaca nobilis Diop_nobi	58	20	1	1	0	0

In Venezuela *Diopsittaca nobilis* is restricted to the Guyana region south of the Orinoco river, and was expected in 361 sampling points from 'San Tomé', Anzoátegui-Monagas state (NeoMaps route NM14); 'Isla de Guara', Monagas-Delta state (NeoMaps route NM36); 'Guri', Bolívar state (NeoMaps route NM24); 'El Manteco', Bolívar state (NeoMaps route NM22); 'Anacoco', Bolívar state (NeoMaps route NM93); 'La Escalera', Bolívar state (NeoMaps route NM90); 'Kavanayén', Bolívar state (NeoMaps route NM21); 'Paraytepuy', Bolívar state (NeoMaps route NM23).



It was detected twice during the field work in 2010, and only has few GBIF record from the 2008-2012 period.

Due to the scarce data available, the model has a poor fit with a combination of high probability of occurrence and low probability of detection and too wide standard errors for the estimated parameters:

```
Call:
occu(formula = ~1 ~ 1, data = UMF[os, ])
Occupancy (logit-scale):
            SE
 Estimate
                   z P(|z|)
     2.82 48.7 0.058
                       0.954
Detection (logit-scale):
 Estimate
            SE
                   z P(|z|)
    -5.82 2.83 -2.06 0.0396
AIC: 31.52855
Number of sites: 450
optim convergence code: 0
optim iterations: 111
Bootstrap iterations: 0
```

3.8 Genus Eupsittula, species Eupsittula pertinax

The genus Eupsittula contains one widespread and common species.

		aspp	GBIF	GBIF.2010	NM.M1	NM.L1	NM.L2	NM.L3
Eupsittula	pertinax	Arat_pert	1102	482	155	27	11	11

Eupsittula pertinax is widely distributed in Venezuela, and was expected in 1180 sampling points.



Available data allowed to fit several alternative models, and models with both climatic and vegetation covariates received most support according to AIC weights.

	spp	mod	n	dtt	AICc	Delta.AICc	AICw	LL
31	Arat_pert	p(.)Psi(VC)	1300	190	1177.645	0.00	0.665	-578.74
32	Arat_pert	p(h)Psi(VC)	1300	190	1179.135	1.49	0.316	-578.46
33	Arat_pert	p(.)Psi(C)	1300	190	1185.397	7.75	0.014	-584.64
34	Arat_pert	p(h)Psi(C)	1300	190	1187.145	9.50	0.006	-584.50
35	Arat_pert	p(.)Psi(V)	1300	190	1284.377	106.73	0.000	-638.17
36	Arat_pert	p(h)Psi(V)	1300	190	1286.388	108.74	0.000	-638.17
37	Arat_pert	nulo	1300	190	1293.649	116.00	0.000	-644.82
38	Arat_pert	p(h)Psi(.)	1300	190	1295.622	117.98	0.000	-644.80

There seems to be a slight effect of the hour of the day on the probability of detection for this species.



Hour of the day

However, the best supported model includes a constant p.

```
Call:
occu(formula = ~1 ~ evi01 + I(evi01^2) + pet01 + I(pet01^2) +
    dT01 + I(dT01<sup>2</sup>) + pre01 + I(pre01<sup>2</sup>), data = UMF[os, ])
Occupancy (logit-scale):
                          SE
                                   z P(|z|)
            Estimate
(Intercept)
             -1.5567 0.2604 -5.979 2.25e-09
evi01
               0.5476 0.1828
                              2.996 2.74e-03
I(evi01^2)
               0.2696 0.1410
                              1.912 5.58e-02
pet01
               0.0308 0.1473
                              0.209 8.34e-01
I(pet01^2)
             -0.1484 0.0713 -2.082 3.74e-02
dT01
               1.8196 0.2364
                              7.697 1.39e-14
I(dT01^2)
               0.1373 0.1712
                              0.802 4.23e-01
               0.0623 0.1910 0.326 7.44e-01
pre01
```

```
I(pre01^2) 0.0602 0.0872 0.691 4.89e-01
Detection (logit-scale):
   Estimate SE z P(>|z|)
   -0.592 0.17 -3.49 0.000485
AIC: 1177.474
Number of sites: 1300
optim convergence code: 0
optim iterations: 51
Bootstrap iterations: 0
```

The following map shows the predicted (unconditional) probability of presence for the whole country based on the model with highest support and the values of the vegetation and climatic covariates.



3.9 Genus Forpus

The genus Forpus is represented in Venezuela by three species, but only one species was detected in NeoMaps surveys of 2010.

		aspp	GBIF	GBIF.2010	NM.M1	NM.L1	NM.L2	NM.L3
Forpus	conspicillatus	Forp_cons	1	0	NA	NA	NA	NA
Forpus	modestus	Forp_mode	9	0	NA	NA	NA	NA
Forpus	passerinus	Forp_pass	991	443	31	8	5	5

3.9.1 Forpus conspicillatus

Forpus conspicillatus is only found in south-western Venezuela, and it was not expected in NeoMaps sampling region.



3.9.2 Forpus modestus

Forpus modestus was expected in 102 sampling points from 'Guri', Bolívar state (NeoMaps route NM24); 'Anacoco', Bolívar state (NeoMaps route NM93). However this species was not detected during the field work in 2010.



3.9.3 Forpus passerinus

Forpus passerinus has a widespread distribution in northern Venezuela. It was expected in 1060 sampling points from NeoMaps sampling localities.



Available data allowed to fit several alternative models, but the model with most support included climatic covariates.

	spp	mod	n	dtt	AICc	Delta.AICc	AICw	LL
49	Forp_pass	p(h)Psi(C)	1050	47	399.3574	0.00	1	-190.59
50	Forp_pass	nulo	1050	47	441.3997	42.04	0	-218.69
51	Forp_pass	p(h)Psi(.)	1050	47	442.0069	42.65	0	-217.99
52	Forp_pass	p(.)Psi(V)	1050	47	443.3331	43.98	0	-217.65
53	Forp_pass	p(h)Psi(V)	1050	47	444.0884	44.73	0	-217.02

The best model includes an effect of time of day on the detection probability.



Hour of the day

The fitted model had high estimated parameter values.

```
Call:
occu(formula = ~hora ~ pet01 + I(pet01^2) + dT01 + I(dT01^2) +
```

```
pre01 + I(pre01^2), data = UMF[os, ])
Occupancy (logit-scale):
            Estimate
                        SE
                                 z P(|z|)
(Intercept)
               1.314 1.256
                                   0.2954
                            1.046
pet01
              -3.515 2.046 -1.718
                                   0.0858
I(pet01^2)
               6.256 2.765 2.262
                                   0.0237
dT01
              -0.652 0.964 -0.677
                                   0.4985
I(dT01^2)
              -4.043 1.750 -2.310 0.0209
              -4.137 2.156 -1.919
pre01
                                  0.0550
              -5.171 2.361 -2.190 0.0285
I(pre01^2)
Detection (logit-scale):
                        SE
            Estimate
                                 z P(>|z|)
(Intercept)
               -2.08 0.789 -2.634 0.00845
               -1.49 2.220 -0.669 0.50339
hora
AIC: 399.1843
Number of sites: 1050
optim convergence code: 0
optim iterations: 63
Bootstrap iterations: 0
```

However, the combination of coefficients resulted in predictions with high contrast in unconditional probability of presence.



3.10 Genus Hapalopsittaca, species H. amazonina

The genus Hapalopsittaca is represented in Venezuela by a single species.

			aspp	GBIF	GBIF	.2010	NM.M1	NM.L1	NM.L2	NM.L3
Hapalopsittaca	amazonina	Hapa_	amaz	12		1	NA	NA	NA	NA
				34						

Hapalopsittaca amazonina is a species of restricted distribution in Venezuela. Hapalopsittaca amazonina was expected in 43 sampling points from 'Jají', Mérida state (NeoMaps route NM26); 'Piñango', Mérida state (NeoMaps route NM57). It was not detected by NeoMaps field work, and only has few GBIF record from the 2008-2012 period.



3.11 Genus Nannopsittaca, species N. panychlora

The genus Nannopsittaca is represented in Venezuela by a single species.

		aspp	GBIF	GBIF.2010	NM.M1	NM.L1	NM.L2	NM.L3
Nannopsittaca	panychlora	Nann_pany	42	15	1	0	0	0

Nannopsittaca panychlora is a species of restricted distribution in Venezuela, it was expected in 146 sampling points from 'La Escalera', Bolívar state (NeoMaps route NM90); 'Kavanayén', Bolívar state (NeoMaps route NM21); 'Paraytepuy', Bolívar state (NeoMaps route NM23).



It was only detected once during NeoMaps field work. The fitted model had large estimates and large uncertainty in both parameters.

```
Call:
occu(formula = ~1 ~ 1, data = UMF[os, ])
Occupancy (logit-scale):
                    z P(|z|)
            SE
 Estimate
     4.41 93.2 0.0473
                        0.962
Detection (logit-scale):
 Estimate SE
                  z P(|z|)
    -5.46 1.5 -3.64 0.000268
AIC: 16.95734
Number of sites: 150
optim convergence code: 0
optim iterations: 73
Bootstrap iterations: 0
```

3.12 Genus Orthopsittaca, species O. manilatus

The genus Orthopsittaca is represented in Venezuela by a single species.

		aspp	GBIF	GBIF.2010	NM.M1	NM.L1	NM.L2	NM.L3
Orthopsittaca	manilatus	Orth_mani	1	0	1	1	2	1

Orthopsittaca manilatus is a species of restricted distribution in Venezuela, it was expected in 407 sampling points from 'San Tomé', Anzoátegui-Monagas state (NeoMaps route NM14); 'Isla de Guara', Monagas-Delta state (NeoMaps route NM36); 'Guri', Bolívar state (NeoMaps route NM24); 'El Manteco', Bolívar state (NeoMaps route NM22); 'Anacoco', Bolívar state (NeoMaps route NM93); 'La Escalera', Bolívar state (NeoMaps route NM90); 'Kavanayén', Bolívar state (NeoMaps route NM21); 'Paraytepuy', Bolívar state (NeoMaps route NM23).



The small number of detections during NeoMaps field work allowed to fit two alternative models with similar AICc support.

sppmodndttAICcDelta.AICcAICwLL55Orth_manip(h)Psi(.)450563.461220.000.522-28.756Orth_maninulo450563.635160.170.478-29.8

The fitted model had large estimates and large uncertainty in both parameters and is thus non-informative for inferences or predictions.

```
Call:
occu(formula = ~hora ~ 1, data = UMF[os, ])
Occupancy (logit-scale):
            SE
 Estimate
                  z P(|z|)
     3.25 40.6 0.08
                      0.936
Detection (logit-scale):
            Estimate
                       SE
                               z P(|z|)
               -1.02 3.26 -0.313
                                    0.754
(Intercept)
              -12.42 9.37 -1.326
hora
                                    0.185
AIC: 63.40741
Number of sites: 450
optim convergence code: 0
optim iterations: 68
```

Bootstrap iterations: 0

3.13 Genus Pionites, species P. melanocephala

The genus Bolborhynchus is represented in Venezuela by a single species.

```
aspp GBIF GBIF.2010 NM.M1 NM.L1 NM.L2 NM.L3
Pionites melanocephala Pion_mela 5 0 3 1 0 1
```

 $Pionites\ melanocephala$ is found in southern Venezuela. It was expected in 0 sampling points from .



With the few number of detections, only one model could be fitted for this species. The fitted model had high estimates and high uncertainty in detectability parameters, and its predictions are not informative.

```
Call:
occu(formula = ~hora ~ 1, data = UMF[os, ])
Occupancy (logit-scale):
 Estimate
            SE
                   z P(|z|)
    -2.52 2.12 -1.19
                     0.234
Detection (logit-scale):
            Estimate
                        SE
                                z P(|z|)
                2.96 5.29 0.559
                                    0.576
(Intercept)
              -15.40 11.96 -1.288
hora
                                    0.198
AIC: 60.56913
Number of sites: 350
optim convergence code: 0
optim iterations: 83
```

3.14 Genus Pionus

Bootstrap iterations: 0

The genus Pionus is represented in Venezuela by five species, but only one species was detected in NeoMaps surveys of 2010.

		aspp	GBIF	GBIF.2010	NM.M1	NM.L1	NM.L2	NM.L3
Pionus	chalcopterus	Pion_chal	44	24	NA	NA	NA	NA
Pionus	fuscus	Pion_fusc	46	4	NA	NA	NA	NA
Pionus	menstruus	Pion_mens	356	86	28	11	7	3

Pionus	seniloides	Pion_seni	0	0	NA	NA	NA	NA
Pionus	sordidus	Pion_sord	142	38	NA	NA	NA	NA

3.14.1 Pionus chalcopterus

Pionus chalcopterus was expected in 38 sampling points from 'Rosario de Perijá', Zulia state (NeoMaps route NM05); 'Sur del Lago', Zulia-Mérida state (NeoMaps route NM07); 'Jají', Mérida state (NeoMaps route NM26). However this species was not detected during the field work in 2010.



3.14.2 Pionus fuscus

Pionus fuscus was expected in 283 sampling points from 'Rosario de Perijá', Zulia state (NeoMaps route NM05); 'Guri', Bolívar state (NeoMaps route NM24); 'El Manteco', Bolívar state (NeoMaps route NM22); 'Caicara del Orinoco', Bolívar state (NeoMaps route NM75); 'Anacoco', Bolívar state (NeoMaps route NM93); 'La Escalera', Bolívar state (NeoMaps route NM90); 'Kavanayén', Bolívar state (NeoMaps route NM21); 'Paraytepuy', Bolívar state (NeoMaps route NM23). However this species was not detected during the field work in 2010.



3.14.3 Pionus menstruus

Pionus menstruus has a widespread but non continuos distribution and was expected in 798 sampling points from NeoMaps.



Available data allowed to fit several alternative models.

	spp	mod	n	dtt	AICc	Delta.AICc	AICw	LL
60	Pion_mens	p(h)Psi(VC)	850	47	361.1700	0.00	0.783	-169.43
61	Pion_mens	p(h)Psi(.)	850	47	364.3834	3.21	0.157	-179.18
62	Pion_mens	p(h)Psi(V)	850	47	366.2987	5.13	0.060	-178.11
63	Pion_mens	p(.)Psi(VC)	850	47	401.6749	40.50	0.000	-190.71
64	Pion_mens	p(.)Psi(V)	850	47	415.4491	54.28	0.000	-203.70
65	Pion_mens	nulo	850	47	419.0509	57.88	0.000	-207.52

The model with most supports include a effect of time of the day on probability of detection and effect of climatic and vegetation covariates on probability of presence.

```
Call:
occu(formula = "hora" evi01 + I(evi01^2) + pet01 + I(pet01^2) +
    dT01 + I(dT01<sup>2</sup>) + pre01 + I(pre01<sup>2</sup>), data = UMF[os, ])
Occupancy (logit-scale):
            Estimate
                        SE
                                z P(|z|)
(Intercept)
              -1.600 1.261 -1.27
                                    0.204
              -1.226 1.038 -1.18
evi01
                                    0.238
I(evi01^2)
               1.873 1.586 1.18
                                    0.238
                                    0.299
               1.463 1.409
                            1.04
pet01
               0.589 0.585 1.01
I(pet01^2)
                                    0.315
dT01
              -2.077 1.587 -1.31
                                    0.191
I(dT01^2)
              -1.102 1.033 -1.07
                                    0.286
pre01
               2.618 1.614 1.62
                                    0.105
              -0.584 0.436 -1.34
I(pre01^2)
                                    0.181
Detection (logit-scale):
            Estimate
                       SE
                               z P(>|z|)
                3.82 1.14 3.36 7.87e-04
(Intercept)
hora
              -19.76 3.63 -5.45 5.04e-08
AIC: 360.8549
Number of sites: 850
optim convergence code: 0
optim iterations: 105
Bootstrap iterations: 0
```

The predicted unconditional probability of presence matches very well the expected distribution.



3.14.4 Pionus seniloides

Pionus seniloides was expected in 128 sampling points from 'Sur del Lago', Zulia-Mérida state (NeoMaps route NM07); 'Jají', Mérida state (NeoMaps route NM26); 'Piñango', Mérida state (NeoMaps route NM57); 'San Joaquín de Navay', Táchira state (NeoMaps route NM18). However this species was not detected during the field work in 2010.



3.14.5 Pionus sordidus

Pionus sordidus was expected in 162 sampling points from 'Rosario de Perijá', Zulia state (NeoMaps route NM05); 'Quebrada Arriba', Lara-Falcón state (NeoMaps route NM03); 'Araya', Sucre state (NeoMaps route NM09); 'Río Tocuyo', Lara state (NeoMaps route NM01); 'Colonia Tovar', Aragua-Vargas state (NeoMaps route NM80). However this species was not detected during the field work in 2010.



3.15 Genus Thectocercus, species T. acuticaudatus

The genus Thectocercus is represented in Venezuela by a single species.

		aspp	GBIF	GBIF.2010	NM.M1	NM.L1	NM.L2	NM.L3
Thectocercus	acuticaudatus	Arat_acut	38	5	3	0	0	0

The ctocercus acuticaudatus was expected in 575 sampling points. Although the species was detected during NeoMaps surveys of 2010, no model could be fitted with the available data.



3.16 Genus Psittacara

The genus Psittacara is represented in Venezuela by two species, both were detected in NeoMaps surveys.

		aspp	GBIF	GBIF.2010	NM.M1	NM.L1	NM.L2	NM.L3
Psittacara	leucophthalmus	Arat_leuc	0	0	5	2	0	0
Psittacara	wagleri	Arat_wagl	1190	101	4	0	0	0

3.16.1 Psittacara leucophthalmus

Psittacara leucophthalmus is only found in eastern Venezuela, and was expected in 411 sampling points from 'Paria', Sucre state (NeoMaps route NM45); 'San Tomé', Anzoátegui-Monagas state (NeoMaps route NM14); 'Isla de Guara', Monagas-Delta state (NeoMaps route NM36); 'Guri', Bolívar state (NeoMaps route NM24); 'El Manteco', Bolívar state (NeoMaps route NM22); 'Caicara del Orinoco', Bolívar state (NeoMaps route NM75); 'Anacoco', Bolívar state (NeoMaps route NM93); 'La Escalera', Bolívar state (NeoMaps route NM90); 'Paraytepuy', Bolívar state (NeoMaps route NM23).



This species was only detected in few NeoMaps localities. Due to the scarce data available, only two models could be fitted. These models had a poor fit with large estimates and large uncertainty for both parameters.

```
AICc Delta.AICc AICw
                                                                LL
         spp
                    mod
                          n dtt
29 Arat_leuc p(h)Psi(.) 450
                                                0.00 0.712 -37.41
                               7 80.87169
                                                1.81 0.288 -39.33
30 Arat_leuc
                   nulo 450
                               7 82.68527
Call:
occu(formula = ~hora ~ 1, data = UMF[os, ])
Occupancy (logit-scale):
 Estimate
            SE
                   z P(|z|)
    -2.75 1.09 -2.51
                       0.012
Detection (logit-scale):
            Estimate
                        SE
                                 z P(|z|)
                3.87 4.04 0.958
(Intercept)
                                     0.338
              -16.50 10.27 -1.607
                                     0.108
hora
AIC: 80.81787
Number of sites: 450
optim convergence code: 0
optim iterations: 44
Bootstrap iterations: 0
```

3.16.2 Psittacara wagleri

Psittacara wagleri is only found in northern mountain ranges of Venezuela, and was expected in 215 sampling points from 'Sur del Lago', Zulia-Mérida state (NeoMaps route NM07); 'Jají', Mérida state (NeoMaps route NM26); 'Piñango', Mérida state (NeoMaps route NM57); 'Yacambú', Lara state (NeoMaps route NM25); 'Colonia Tovar', Aragua-Vargas state (NeoMaps route NM80).



This species was only detected in few NeoMaps localities. Due to the scarce data available only one model could be fitted. This model had a poor fit with a combination of low probability of occurrence and high probability of detection and too wide standard errors for the estimated parameters:

AICc Delta.AICc AICw LL mod n dtt spp 39 Arat_wagl p(h)Psi(.) 250 4 47.11455 0 1 -20.51 Call: occu(formula = ~hora ~ 1, data = UMF[os,]) Occupancy (logit-scale): Estimate SE z P(>|z|)-4.12 0.504 -8.17 3.04e-16 Detection (logit-scale): Estimate z P(>|z|) SE 12.24 452 0.0271 0.978 (Intercept) hora -1.41 NaN NaN NaN AIC: 47.01699 Number of sites: 250 optim convergence code: 0 optim iterations: 23 Bootstrap iterations: 0

3.17 Genus Pyrilia

The genus Pyrilia is represented in Venezuela by three species, only one species was detected in NeoMaps surveys.

		aspp	GBIF	GBIF.2010	NM.M1	NM.L1	NM.L2	NM.L3
Pyrilia	barrabandi	Pion_barr	36	11	0	2	0	0
Pyrilia	caica	Pyri_caic	26	4	NA	NA	NA	NA
Pyrilia	pyrilia	Pyri_pyri	20	12	NA	NA	NA	NA

3.17.1 Pyrilia barrabandi

Pyrilia barrabandi is only found in southern Venezuela, and was expected in 50 sampling points from 'Gavilán', Amazonas state (NeoMaps route NM34).



It was detected twice during the field work in 2010. Due to the scarce data available, only two models could be fitted. The null model had best support according to AIC weights, but a poor fit with a combination of high probability of occurrence and low probability of detection and too wide standard errors for the estimated parameters:

```
LL
                        n dtt
                                    AICc Delta.AICc AICw
         spp
                    mod
57 Pion_barr
                   nulo 50
                              2 22.75325
                                               0.00 0.744 -9.25
58 Pion_barr p(h)Psi(.) 50
                             2 24.89061
                                               2.14 0.256 -9.18
Call:
occu(formula = ~1 ~ 1, data = UMF[os, ])
Occupancy (logit-scale):
 Estimate
            SE
                    z P(|z|)
     3.98 53.9 0.0738
                        0.941
Detection (logit-scale):
            SE
                   z P(|z|)
 Estimate
```

```
-3.59 1.25 -2.86 0.00417
AIC: 22.49793
Number of sites: 50
optim convergence code: 0
optim iterations: 69
Bootstrap iterations: 0
```

3.17.2 Pyrilia caica

Pyrilia caica was expected in 299 sampling points from 'Guri', Bolívar state (NeoMaps route NM24); 'El Manteco', Bolívar state (NeoMaps route NM22); 'Anacoco', Bolívar state (NeoMaps route NM93); 'La Escalera', Bolívar state (NeoMaps route NM90); 'Kavanayén', Bolívar state (NeoMaps route NM21); 'Paraytepuy', Bolívar state (NeoMaps route NM23). However this species was not detected during the field work in 2010.



3.17.3 Pyrilia pyrilia

Pyrilia pyrilia was expected in 150 sampling points from 'Quebrada Arriba', Lara-Falcón state (NeoMaps route NM03); 'Sur del Lago', Zulia-Mérida state (NeoMaps route NM07); 'Jají', Mérida state (NeoMaps route NM26); 'Capitanejo', Barinas state (NeoMaps route NM19); 'Otopún', Barinas state (NeoMaps route NM16); 'San Joaquín de Navay', Táchira state (NeoMaps route NM18). However this species was not detected during the field work in 2010.



3.18 Genus Pyrrhura

The genus Pyrrhura is represented in Venezuela by seven species and five were detected in NeoMaps surveys.

		aspp	GBIF	GBIF.2010	NM.M1	NM.L1	NM.L2	NM.L3
Pyrrhura	caeruleiceps	Pyrr_caer	37	2	NA	NA	NA	NA
Pyrrhura	egregia	Pyrr_egre	60	9	3	1	0	0
Pyrrhura	emma	Pyrr_emma	37	2	NA	NA	NA	NA
Pyrrhura	hoematotis	Pyrr_hoem	234	71	1	0	0	0
Pyrrhura	melanura	Pyrr_mela	10	3	1	1	0	1
Pyrrhura	picta	Pyrr_pict	130	31	6	2	2	0
Pyrrhura	rhodocephala	Pyrr_rhod	96	23	1	0	0	0

3.18.1 Pyrrhura caeruleiceps

Pyrrhura caeruleiceps was expected in 7 sampling points from 'Rosario de Perijá', Zulia state (NeoMaps route NM05). However this species was not detected during the field work in 2010.



3.18.2 Pyrrhura egregia

Pyrrhura egregia was expected in 152 sampling points from 'La Escalera', Bolívar state (NeoMaps route NM90); 'Kavanayén', Bolívar state (NeoMaps route NM21); 'Paraytepuy', Bolívar state (NeoMaps route NM23).



This species was detected in few localities during the field work in 2010. Several alternative models were fitted.

	spp	mod	n	dtt	AICc	Delta.AICc	AICw	LL
66	Pyrr_egre	p(.)Psi(V)	150	3	32.51359	0.00	0.460	-12.12
67	Pyrr_egre	nulo	150	3	33.49337	0.98	0.282	-14.71
68	Pyrr_egre	p(h)Psi(V)	150	3	34.65420	2.14	0.158	-12.12
69	Pyrr_egre	p(h)Psi(.)	150	3	35.57612	3.06	0.100	-14.71

The fitted models with extreme values in parameters and high uncertainty in estimates are not informative.

```
Call:
occu(formula = ~1 ~ evi01 + I(evi01^2), data = UMF[os, ])
Occupancy (logit-scale):
                        SE
                                 z P(|z|)
            Estimate
               -3.40 0.899 -3.784 0.000154
(Intercept)
evi01
                2.07 1.828
                            1.131 0.258179
I(evi01^2)
               -1.44 2.180 -0.661 0.508699
Detection (logit-scale):
            SE
 Estimate
                  z P(|z|)
     9.44 85.4 0.11
                      0.912
AIC: 32.23772
Number of sites: 150
optim convergence code: 0
optim iterations: 63
Bootstrap iterations: 0
```

3.18.3 Pyrrhura emma

Pyrrhura emma was expected in 107 sampling points from 'Araya', Sucre state (NeoMaps route NM09); 'Paria', Sucre state (NeoMaps route NM45); 'Colonia Tovar', Aragua-Vargas state (NeoMaps route NM80). However this species was not detected during the field work in 2010.



3.18.4 Pyrrhura hoematotis

Pyrrhura hoematotis was expected in 85 sampling points from 'Yacambú', Lara state (NeoMaps route NM25); 'Colonia Tovar', Aragua-Vargas state (NeoMaps route NM80).



This species was only detected once during the field work in 2010.

		spp	$\verb+mod$	n	dtt	AICc	Delta.AICc	AICw	LL
70	Pyrr_	hoem	nulo	100	1	15.32454	0	1	-5.6

The fitted models with extreme values in parameters and high uncertainty in estimates are not informative.

```
Call:
occu(formula = ~1 ~ 1, data = UMF[os, ])
Occupancy (logit-scale):
 Estimate
            SE
                   z P(|z|)
    -4.59 1.01 -4.56 5.07e-06
Detection (logit-scale):
 Estimate SE
                 z P(|z|)
     6.66 62 0.107
                     0.915
AIC: 15.20083
Number of sites: 100
optim convergence code: 0
optim iterations: 66
Bootstrap iterations: 0
```

3.18.5 Pyrrhura melanura

Pyrrhura melanura was expected in 50 sampling points from 'Gavilán', Amazonas state (NeoMaps route NM34).



This species was only detected in few occasions during the field work in 2010.

	spp	mod	n	dtt	AICc	Delta.AICc	AICw	LL
71	Pyrr_mela	p(h)Psi(.)	50	3	31.23506	0	1	-12.36

The fitted model with extreme values in parameters and high uncertainty in estimates are not informative.

```
Call:
occu(formula = ~hora ~ 1, data = UMF[os, ])
Occupancy (logit-scale):
 Estimate
            SE
                    z P(|z|)
     6.16 91.5 0.0673
                        0.946
Detection (logit-scale):
            Estimate
                        SE
                                z P(|z|)
              -0.602 3.19 -0.189
                                     0.850
(Intercept)
hora
              -8.094 10.12 -0.800
                                     0.424
AIC: 30.71332
Number of sites: 50
optim convergence code: 0
optim iterations: 40
Bootstrap iterations: 0
```

3.18.6 Pyrrhura picta

Pyrrhura picta was expected in 317 sampling points from 'Isla de Guara', Monagas-Delta state (NeoMaps route NM36); 'Guri', Bolívar state (NeoMaps route NM24); 'El Manteco', Bolívar state (NeoMaps route NM22); 'Anacoco', Bolívar state (NeoMaps route NM93); 'La

Escalera', Bolívar state (NeoMaps route NM90); 'Kavanayén', Bolívar state (NeoMaps route NM21); 'Paraytepuy', Bolívar state (NeoMaps route NM23).



Available data allowed to fit several alternatic models.

	spp	mod	n	dtt	AICc	Delta.AICc	AICw	LL
72	Pyrr_pict	p(.)Psi(V)	350	10	89.88223	0.00	0.724	-40.88
73	Pyrr_pict	p(h)Psi(V)	350	10	91.82109	1.94	0.275	-40.82
74	Pyrr_pict	nulo	350	10	104.14575	14.26	0.001	-50.06
75	Pyrr_pict	p(h)Psi(.)	350	10	106.01953	16.14	0.000	-49.98

According to the model with the best support the unconditional probability of presence varied along the NeoMaps routes depending on the value of the covariates. The figure show four examples of 40 km routes, ordered from beginning to end, three with detections (NM22, NM36 and NM93) and one without detections (NM24). The lines show the predicted unconditional probability of presence (solid line: best estimate, dotted lines 95% confidence interval), and the dots represent the conditional probability given the observed detection history. Blue dots represent the localities of known occurrences, red dots are localities with very low posterior probabilities of presence ($\Psi_{post} < 0.125$), grey dots are localities with intermediate values.



3.18.7 Pyrrhura rhodocephala

Pyrrhura rhodocephala was expected in 160 sampling points from 'Sur del Lago', Zulia-Mérida state (NeoMaps route NM07); 'Jají', Mérida state (NeoMaps route NM26); 'Piñango', Mérida state (NeoMaps route NM57); 'Yacambú', Lara state (NeoMaps route NM25).



This species was only detected once during the field work in 2010.

	spp	mod	n	dtt	AICc	Delta.AICc	AICw	LL
76	Pyrr_rhod	nulo	200	1	16.65263	0.00	0.737	-6.3
77	Pyrr_rhod	p(h)Psi(.)	200	1	18.71410	2.06	0.263	-6.3

The fitted models with extreme values in parameters and high uncertainty in estimates are not informative.

```
Call:
occu(formula = ~1 ~ 1, data = UMF[os, ])
Occupancy (logit-scale):
Estimate SE z P(>|z|)
   -5.29 1 -5.28 1.31e-07
Detection (logit-scale):
Estimate SE z P(>|z|)
    8.41 152 0.0555 0.956
AIC: 16.59171
Number of sites: 200
optim convergence code: 0
optim iterations: 15
Bootstrap iterations: 0
```

3.19 Genus Touit

No species of the genus Touit was detected by NM surveys. There are also few records in GBIF.

		i	aspp	GBIF	GBIF.2010	NM.M1	NM.L1	NM.L2	NM.L3
Touit	batavica	Toui_	bata	1	0	NA	NA	NA	NA
Touit	dilectissima	Toui_	dile	0	0	NA	NA	NA	NA
Touit	huetii	Toui_	huet	4	2	NA	NA	NA	NA
Touit	purpurata	Toui_]	purp	0	0	NA	NA	NA	NA

3.19.1 Touit batavica

This species was expected in 183 NM sampling localities.



3.19.2 Touit dilectissima

This species was expected in 87 NM sampling localities.



3.19.3 Touit huetii

This species was expected in 213 NM sampling localities.



3.19.4 Touit purpurata

This species was expected in 222 NM sampling localities.



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