Peer Community Journal

Section: Evolutionary Biology

RESEARCH ARTICLE

Published
2021-11-24

Cite as
Amélie Vantaux, Franck Yao, Domonbabele FdS Hien, Edwige Guissou, Bienvenue K. Yameogo, Louis-Clément Gouagna, Didier Fontenille, François Renaud, Frédéric Simard, Carlo Constantini, Fréderic Thomas, Karine Mouline, Benjamin Roche, Anna Cohuet, Kounbobr R Dabiré and Thierry Lefèvre (2021) *Field evidence for manipulation of mosquito host selection by the human malaria parasite*, Plasmodium falciparum, Peer Community Journal, 1: e13.

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Peer-review
Peer reviewed and recommended by PCI Evolutionary Biology, https://doi.org/10.24072/pci.evolbiol.100057

Field evidence for manipulation of mosquito host selection by the human malaria parasite, *Plasmodium falciparum*

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Volume **1** (2021), article e13

https://doi.org/10.24072/pcjournal.13

Abstract

Whether the malaria parasite Plasmodium falciparum can manipulate mosquito host choice in ways that enhance parasite transmission toward humans is unknown. We assessed the influence of P. falciparum on the blood-feeding behaviour of three of its major vectors (Anopheles coluzzii, An. gambiae and An. arabiensis) in Burkina Faso. Host preference assays using odour-baited traps revealed no effect of infection on mosquito long-range anthropophily. However, the identification of the blood meal origin of mosquitoes showed that females carrying sporozoites, the mature transmissible stage of the parasite, displayed a 24% increase in anthropophagy compared to both females harbouring oocysts, the parasite immature stage, and uninfected individuals. Using a mathematical model, we further showed that this increased anthropophagy in infectious females resulted in a 250% increase in parasite transmission potential, everything else being equal. This important epidemiological consequence highlights the importance of vector control tools targeting infectious females.

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Introduction

There is mounting evidence that malaria parasites affect phenotypic traits of their vectors and hosts in ways that increase contacts between them, hence favouring parasite transmission (Hurd, 2003; Koella, 2005; Lefèvre & Thomas, 2008). In addition to increased vertebrate attractiveness to mosquito vectors (Batista et al., 2014; Busula et al., 2017; Cornet et al., 2013; De Moraes et al., 2014; Emami et al., 2017; Lacroix et al., 2005), another frequently reported parasite-induced change is the alteration of vector motivation and avidity to feed (L. J. Cator et al., 2012; Stanczyk et al., 2017). Mosquitoes infected with *Plasmodium* sporozoites (the mosquito to human transmission stage) can indeed display increased (i) responses to host odours (L. J. Cator et al., 2013; Rossignol et al., 1986), (ii) landing and biting activity (Anderson et al., 1999; Koella et al., 2002; Rossignol et al., 1984, 1986; Smallegange et al., 2013; Wekesa et al., 1992), (iii) number of feeds (Koella et al., 1998) and (iv) blood volume intake (Koella & Packer, 1996; Koella et al., 2002; Koella et al., 1998). In contrast, mosquitoes infected with oocysts (the immature non-transmissible stage of the parasite), are less likely to attempt to feed (Anderson et al., 1999; L. J. Cator et al., 2013; Koella et al., 2002). Since biting is risky (e.g., host defensive behaviours can kill the vector and its parasite), reduced feeding attempts would be beneficial to the parasite during the non-transmissible stage as this would reduce mortality before the parasite reaches maturity and is ready to be transmitted (Schwartz & Koella, 2001).

These "stage-dependent" behavioural alterations likely increase parasite transmission (L. Cator et al., 2014; Dobson, 1988), provided that mosquito feeds are taken on a suitable vertebrate host species for the parasite. While malaria vectors can usually feed on a range of different vertebrate species (Takken & Verhulst, 2013), the malaria parasites they transmit are often highly host-specific, infecting only one or a few vertebrate species (Perkins, 2014). For example *P. falciparum*, which causes the most severe form of human malaria, displays an extreme form of specificity and can develop and reproduce in hominids only (predominantly in humans and to a lesser extent in chimpanzees, bonobos, and gorillas) (Ngoubangoye et al., 2016; Prugnolle et al., 2011; Rayner et al., 2011), such that any mosquito bite on another vertebrate species would be a dead-end for the parasite. In contrast, the vectors of *P. falciparum* can feed on a wide range of vertebrate host species in the wild depending on the geographic area and the relative abundance of humans and other vertebrates (Costantini et al., 1999; Takken & Verhulst, 2013). Accordingly, *P. falciparum* could modify its vector choice in ways that enhance transmission toward humans and/or reduce mosquito attraction to other unsuitable host species (i.e. specific manipulation). A previous study testing this hypothesis found no effect of *P. falciparum* infection on host preference of three major vector species, *An. coluzzii*, *An. gambiae,* and *An. arabiensis* (Nguyen et al., 2017). However, this study examined the odour-mediated mosquito host preference in laboratory conditions using a dual-port olfactometer, not the final realised host choice which is of primary importance for parasite transmission.

Here, we assessed the influence of *P. falciparum* on *An. coluzzii, An. gambiae* and *An. arabiensis* blood-feeding behaviour in three villages in Burkina Faso. First, odour-baited traps, set side by side in a choice arrangement and releasing either human or calf odours were used to determine odour-mediated mosquito host preference (Experiment 1). Second, indoor-resting blood-fed mosquito females were collected and the origin of their blood meal was identified to determine mosquito host selection (Experiment 2). Third, we quantified the epidemiological consequences of variation in the patterns of host selection using a compartmental model for *Plasmodium* transmission between humans and mosquitoes.

Material and methods

Collection sites

The study was conducted in three villages in South-Western Burkina Faso: Soumousso (11°23'14"N, 4°24'42"W), Klesso (10°56'40.5"N, 3°59'09.9"W) and Samendeni (11°27'14.3"N, 4°27'37.6"W) (Figure supplement S1). The three villages are located in an area characterized by wooded savannah, where *Anopheles* females only have access to temporary, rain-filled puddles and quarries that permit larval development during the rainy season from June to November. The dry season extends from December to May. In these rural villages, domestic animals (including cattle, goats, sheep, pigs, chickens, donkeys, dogs)

are usually kept in compounds in open conditions but a few households use separate roofed shelters for sheep, goats, pigs and chickens. Most houses are mud-walled with roofs of iron sheets or thatch, but a few houses are made of bricks.

Experiment 1: Mosquito host preference

Two odour-baited entry traps (OBETs as in Costantini et al., 1996; Costantini et al., 1998; Lefèvre et al., 2009) and two odour-baited double net traps (BNTs as in Tangena et al., 2015) baited with calf and human odours were used to assess the host preference of field populations of mosquitoes in Samandeni and Klesso villages (Figure 1).The two OBETs were connected to a tent (Lxlxh: 250x150x150 cm) by air vent hoses (Scanpart®, DxL=10*300cm; Figure 1a). The odours of the two hosts were drawn by a 12-V fan from the tents and into the OBETs by the air vent hoses, coming out of the traps at a speed of 15cm/s (±2cm/s), as measured with a Testo 425-Compact Thermal Anemometer (Testo, Forbach, France) equipped with a hot wire probe [range: 0 to + 20m/s, accuracy: ± (0.03 m/s + 5% of mv)]. Host-seeking mosquitoes responding to the host cues flew up the odour-laden streams and entered one of the two traps. The two odour-baited double net traps (BNTs) consisted of an untreated bed net (Lxlxh: 300x250x185 cm) from which each corner was raised 20 cm above ground and a smaller untreated bed net (Lxlxh: 190x120x150 cm) protecting the human volunteer in the human baited trap (Figure 1b).

In both OBETs and BNTs, the human volunteers rested on a metal-framed bed (Lxl: 190x80 cm) and were protected from mosquito bites. OBETs and BNTs were operated from 19:00 to 05:30 hours, for 3 nights in June 2013, and 13 nights in September 2013 in Samendeni. The BDNTs only were set-up for 6 nights in September in Klesso. Different combinations of live calves and humans were used as odour sources on each testing day to obviate any individual effect. Calves of about similar size and weight as human volunteers were used to equalize the quantity of emitted odours. Trapped mosquitoes were retrieved in the morning using mouth aspirators. They were kept in a 20x20x20 cm cage with a humid towel on top and brought back to the laboratory for further processing (see below).

Experiment 2: Mosquito blood-feeding pattern

Indoor resting blood-fed mosquitoes were collected between 7 am and 9 am by insecticide spray catches as in Lefèvre et al. (2009) to determine the origin of their blood-meal. Briefly, white sheets were spread over the floor surface and the furniture inside houses. The houses were then sprayed with an insecticide (Kaltox®: allethrin 0.27%, tetramethrin 0.20 %, permethrin 0.17%, propoxur 0.68%) to knock down the mosquitoes. Fifteen minutes after spraying, blood-fed *An. gambiae s.l.* mosquitoes were collected from the white sheet using forceps and placed on moist filter paper inside labeled petri dishes.

In Samandeni and Klesso, mosquito collections were carried out in the rainy season only (4 days in June 2013, and 13 days in September 2013 in Samendeni, and 6 days in September 2015 in Klesso), whereas in Soumousso they were conducted in both the rainy and the dry season (26 days between January and November 2009). In Soumousso, human dwellings (from 10 neighbourhoods) only were sampled whereas animal sheds and unoccupied houses were also sampled in Samandeni and Klesso. A total of 27 human dwellings, 7 unoccupied houses and 20 animal sheds were sampled in Samendeni. A total of 7 human dwellings, 7 unoccupied houses and 9 animal sheds were sampled in Klesso. All mosquitoes were kept in a Petri dish with a humid paper towel to facilitate later dissection and brought back to the laboratory for further processing (see below).

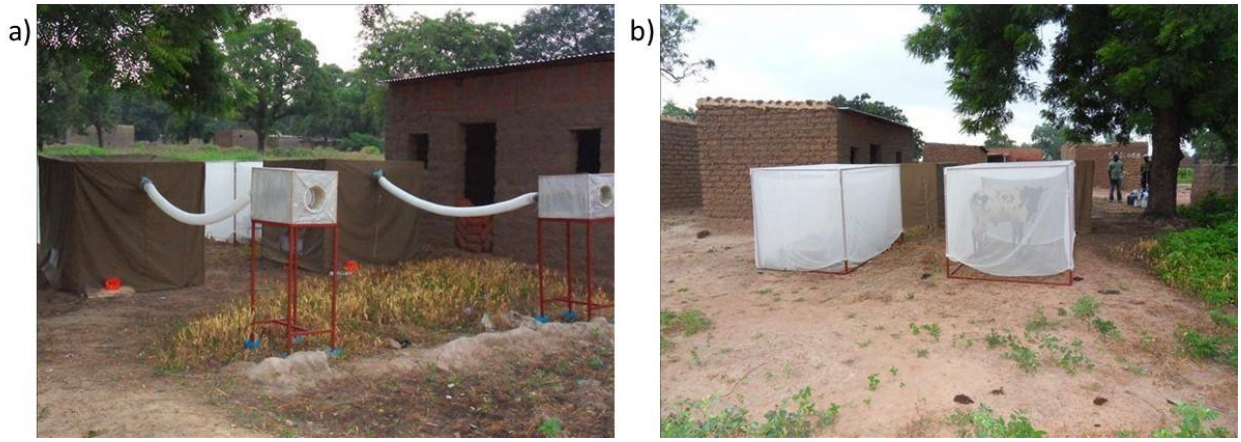

Figure 1. Traps baited with calf and human odours used to assess the host preference of field populations of mosquitoes in Samandeni and Klesso villages. **a)** Two odour-baited entry traps (OBETs) were connected to a tent by air vent hoses. **b)** Two odour-baited double net traps (BNTs).

Laboratory processing of samples

A total of 3447 blood-fed *Anopheles gambiae s.l*. collected indoors (Experiment 2) and 674 females collected in the choice traps (Experiment 1) were processed. In addition, a subset of 276 females collected indoors was used to determine parity (parous versus nulliparous) based on the condition of ovarian tracheoles in order to control for age. Similarly, a subset of 418 individuals was used to determine different species within the *Anopheles gambiae sensu stricto* complex (i.e. distinguishing *Anopheles arabiensis*, *Anopheles coluzzii* and *Anopheles gambiae*) using routine PCR-RFLP based on segregating SNP polymorphisms in the X-linked ribosomal DNA InterGenic Spacer region as described in Santolamazza *et al*. (2008).

Anopheles gambiae sl. females were dissected in a drop of phosphate buffered saline (PBS) (pH 7.2). Blood-fed midguts were gently squeezed under a stereomicroscope (magnification 35x, Leica EZ4D, Wetzlar, Deutschland) to get the blood out, which was mixed with PBS, absorbed on a filter paper, and then kept at -20°C until identification by an enzyme-linked-immunosorbent assay (ELISA) for Soumousso and Samendeni samples (Beier et al., 1988) and by multiplex PCR for Klesso samples (Kent & Norris, 2005). Each blood meal was discriminated between human, cattle, goat/sheep, chicken, dog, pig, and horse/donkey origins. ELISA-based determination of mosquito blood meal origin was performed using anti-human IgG-, anti-bovine IgG-, anti-pig IgG, anti-chicken IgG-, anti-goat IgG-, anti-sheep IgG-, anti-dog IgG-, and anti-horse IgG-peroxidase conjugates (A8794, A5295, A5670, A9046, A5420, A3415, A6792, A6917, Sigma-Aldrich). PCR-based determination of the mosquito blood meal origin targeting the vertebrate host cytochrome B was performed as described by Kent and Norris (2005), with the following modifications: *(i)* Three additional primers were designed from available Genbank sequences to target the following potential hosts: chicken470F (Genbank accession number: AB044986.1), sheep695F (KY662385.1), donkey574F (FJ428520.1); *(ii)* for each individual, two multiplex reactions were performed to avoid cross-reactions between primers and to optimize the determination. In the multiplex reaction #1, UNREV1025, Chicken470F, Sheep695F, Goat894F and Donkey574F primers were used at an amplification temperature of 49.2 °C. In the multiplex reaction #2, UNREV1025, Dog368F, Human741F, Cow121F and Pig573F primers were used at an amplification temperature of 58°C. Blood meal origin diagnostic was based on the PCR products expected sizes as follow: donkey (460bp), sheep (340bp), chicken (290bp), goat (150bp), dog (680bp), cow (561bp), pig (453bp), human (334bp).

The extracted midguts were then stained with 1% Mercurochrome® solution to detect with a microscope (magnification 400x, Leica ICC50, Wetzlar, Deutschland) the presence and number of *Plasmodium* spp. oocysts. PCR on a subset of oocyst-infected individuals (20 midguts of a total of 118 oocyst-infected individuals) confirmed that these oocysts all belonged to *P. falciparum*. The head and thorax of individual mosquitoes were stored at −20°C in 1.5 mL Eppendorf tubes. Sporozoite infection with *P. falciparum* was determined by ELISA using peroxidase-conjugated *Plasmodium falciparum* circumsporozoite protein monoclonal antibody for the Soumousso samples (Wirtz et al., 1987) and by qPCR

for the samples from Samendeni and Klesso (Boissière et al., 2013). The quantification of *P. falciparum* sporozoites in salivary glands was determined by qPCR using 7500 Fast Real time PCR System (Applied Biosystems, Foster City CA, USA). The mosquito heads and thoraxes were crushed individually and DNA extracted as previously described (Morlais et al., 2004). For sporozoite quantification, we targeted the fragment of subunit 1 of the mitochondrial cytochrome c oxidase gene (cox 1) using the forward and reverse primer sequences, qPCR-PfF 5'-TTACATCAGGAATGTTATTGC-3' and qPCR-PfR 5'-ATATTGGATCTCCTGCAAAT-3, respectively. The reaction was conducted in a 10µL final volume containing: 1µL of DNA template, 1x HOT Pol EvaGreen qPCR Mix Plus ROX, and 600nM of each primer. Amplification was started by an initial activation step at 95°C for 15min and 40 cycles of denaturation at 95°C for 15s and annealing / extension at 58°C for 30s. Detection was conducted during the last step (Boissière et al., 2013). Quantification was based on a standard curve built from four serial dilutions (12%) of an asexual parasite culture. We made dilutions ranging from 60 to 60,000 genome/µl of DNAs from a standard culture. The first dilution (10^{-1}) was used as a positive control. The standard curve (y= -3.384X +35.874) was obtained by linear regression analysis of Ct values (Cycle threshold) *versus* log10 genome copy number of parasite culture.

This protocol allowed us to gather the following information for each collected individual mosquito: immature *Plasmodium* infection status (presence of oocysts in the midgut); mature *P. falciparum* infection status (presence of sporozoites in salivary glands); source of blood meal or trap (calf/human) chosen; shelter type (human dwellings, unoccupied houses, animal sheds).

Statistical analyses

Experiment 1: Mosquito host preference -The anthropophily index (AI) was expressed as the number of *Anopheles gambiae s.l.* caught in the human-baited trap over the total number of mosquitoes caught in both human- and calf- baited traps. We tested the effect of infection status (uninfected, infected with the oocyst immature stages and infected with the sporozoite transmissible stages), collection method (OBET *vs*. BNT), and their interaction on AI using a General Linear Model (GLM) with a binomial error structure.

Experiment 2: Mosquito blood-feeding pattern -The human blood index (HBI) was expressed as the number of *Anopheles gambiae s.l.* fed on humans including mixed human-animal blood meals over the total number of blood-fed *Anopheles gambiae s.l.*. We tested the effect of *Plasmodium* infection status (uninfected, oocyst-infected, sporozoite-infected individuals - 25 individuals with both oocysts and sporozoites were included in the sporozoite infected group and excluding these individuals from the analysis yielded similar results), village (Soumousso, Samendeni, Klesso), shelter type (human dwelling, unoccupied house, animal shed) and relevant two-way interactions (infection status by shelter type and infection status by village) on HBI using a GLM with a binomial error structure. The effect of species (*Anopheles gambiae*, *An. coluzzii* and *An. arabiensis*), infection status, shelter type, and their interactions on HBI was assessed using the subset of females identified to the molecular level using a GLM with a binomial error structure. The effect of parity (nulliparous *vs*. parous) on HBI was assessed on a subset of females using a GLM with a binomial error structure.

We also verified for both AI and HBI whether choice significantly differed from a random distribution between humans and animals or whether mosquitoes displayed a statistically significant attraction to one type of blood meal or trap.

For model selection, we used the stepwise removal of terms, followed by likelihood ratio tests (LRT). Term removals that significantly reduced explanatory power ($P<0.05$) were retained in the minimal adequate model (Crawley, 2007). All analyses were performed in R v.3.0.3.

Mathematical model

In order to explore the epidemiological consequences of variation in HBI, we built a compartmental model for *Plasmodium* transmission between humans and mosquitoes (Keeling & Rohani, 2008):

$$\frac{dS_m}{dt} = \mu N_m - ab\frac{S_m}{N_m} I_h \varepsilon_s - \mu S_m$$

$$\frac{dE_m}{dt} = ab\frac{S_m}{N_m} I_h \varepsilon_s - (\mu + \gamma) E_m$$

$$\frac{dI_m}{dt} = \gamma E_m - \mu I_m$$

$$\frac{dS_h}{dt} = -ac\frac{S_h}{N_h} I_m \varepsilon_i + \delta I_h$$

$$\frac{dI_h}{dt} = ac\frac{S_h}{N_h} I_m \varepsilon_i - \delta I_h$$

Susceptible mosquitoes (S_m) are born at rate μ and become exposed (E_m) according to their biting rate (a), their probability to get infected (b) and the HBI of susceptible mosquitoes (ε_s). Then, exposed mosquitoes become infectious (I_m) according to their extrinsic incubation period (γ). Mosquito population die at rate (μ). Nm is the number of mosquitoes. Susceptible humans (S_h) get infected according to mosquito biting rate, the probability to develop infection (c) and the HBI of infectious mosquitoes (ε_i). Nh is the number of humans. Then, infectious humans remain infectious (I_h) during a period equals to $1/\delta$ on average. See parameter values in table supplement S1 (Roux et al., 2015; Vantaux et al., 2016). In our simulation we based the HBI of exposed mosquitoes (ε_s) on the confidence intervals of oocyst-infected mosquitoes that were experimentally measured in this study. Then we explored the impact of the HBI of infectious mosquitoes (ε_i, during the sporozoite stage) on the Entomological Inoculation Rate (EIR), representing the number of infectious bites received by a human during one year (D. Smith & Ellis McKenzie, 2004), as defined by:

$$EIR = ma\frac{I_m}{N_m}$$

where m is the ratio between mosquitoes and humans, and other parameters are as above. We kept an identical human population size of 100 individuals and only varied mosquito densities to assume different ratio values (m) between mosquitoes and humans (low: m=1, medium: m=10 and high: m=100) in order to explore the impact of different HBIs on the EIR in relation to mosquito densities. Then, the mathematical model was simulated for one season in order to estimate the proportion of infectious mosquitoes.

Ethics

Ethical approval was obtained from the Centre Muraz Institutional Ethics Committee under agreement no. 0003-2009/CE-CM and A0003-2012/CE-CM.

Results

Experiment 1: Mosquito host preference

To assess the inherent mosquito host preference of field populations of mosquitoes, we used two odour-baited entry traps (OBETs) and two odour-baited double net traps (BNTs) releasing either calf or human odours. The anthropophily index (AI) was expressed as the number of Anopheles gambiae s.l. caught in the human-baited trap over the total number of mosquitoes caught in both human- and calf-baited traps. The infection status was successfully determined in 584 out of the 674 mosquitoes (86.6%)

collected in the OBETs (383 individuals) and BNTs (201 individuals). Uninfected, oocyst-infected and sporozoite-infected females displayed similar host preferences (X^2_2 = 3.6, P = 0.17, Figure supplement S2, AI uninfected females: 63.3 ± 4%, N=531, OR=0.58, 95% CI = 0.53-0.63, P <0.0001; AI oocyst-infected females: 55.2 ± 18 %, N=29, OR=0.81, 95% CI = 0.56-1.18, P=0.58; AI sporozoite-infected females: 45.8 ± 20 %; N=24, OR=1.18, 95% CI = 0.78-1.78, P=0.7). There was no effect of collection method on AI (OBETs: 64 ± 5%, BNTs: 59 ± 7%; X^2_1 = 1.5, P = 0.21), indicating that both methods are comparable to assess mosquito host preference. There was no interaction between mosquito infection and collection method (X^2_2 = 0.26, P = 0.9; Figure supplement S2).

Experiment 2: Mosquito blood-feeding pattern

To assess the realized host selection of Anopheles gambiae s.l., the blood meal origins of indoor-resting females were identified. The human blood index (HBI) was expressed as the number of females fed on humans (including mixed human-animal blood meals) over the total number of blood-fed females. Of the 3447 blood-fed Anopheles gambiae s.l. collected indoors, the blood meal origin was successfully identified in 2627 samples (76%). Among these 2627 samples, infection status was successfully determined in 2328 mosquitoes (88.6%). The following analyses are restricted to these 2328 females. HBI was significantly affected by mosquito infection status (X^2_2 = 13.007, P = 0.0015; Figure 2) with a 24% increase in HBI in sporozoite-infected females compared to both their oocyst-infected and uninfected counterparts (sporozoite-infected: 77 ± 5.7%; N=209, deviation from random feeding: OR=0.3, 95% CI = 0.25-035, P <0.0001; oocyst-infected females: 63.6 ± 5.7%, N=118, OR=0.57, 95% CI = 0.47-0.69, P =0.004; uninfected females: 61.1 ± 2.1%; N=2001, OR=0.64, 95% CI = 0.61-0.66, P <0.0001). However, because sample size in the uninfected group (N=2001) was higher than that of both sporozoite-infected (N= 209) and oocyst-infected groups (N=118), we ran a second set of analyses using a subset of 150 randomly selected uninfected individuals. This approach normalizes statistical power to test for statistically significant differences in HBI across heterogeneous sample sets. The randomisation was repeated 100 times and the analysis confirmed a significantly higher anthropophagy in sporozoite-infected individuals compared to both oocyst-infected individuals and uninfected individuals in 100% of these randomisations (mean (X^2_2) = 12.7, CI (X^2_2) = (7.54-21.59), mean (P) = 0.0043, CI(P) = (0.00002-0.023); Tukey post-hoc tests: sporozoite-infected vs. oocyst-infected individuals, this pair-wise comparison was significantly different in 100 % of the randomisations: mean(P) = 0.02577, CI(P) = (0.02559-0.02591); sporozoite-infected vs. uninfected individuals, this pair-wise comparison was significantly different in 90% of the randomisations: mean (P) = 0.023, CI(P) = (5e-07 - 3e-01); oocyst-infected vs. uninfected individuals, this pair-wise comparison was significantly different in 0 % of the randomisations: mean (P) = 0.78, CI(P) = (0.07-0.99)).

The HBI of sporozoite-infected mosquitoes was higher than that of oocyst-infected and uninfected females regardless of the village considered (infection status: village interaction: X24 = 2.3, P = 0.68, Figure 2) or the shelter type in which mosquito females were collected (infection status: shelter type interaction: X24 = 0.7, P = 0.95, Figure supplement S3).

HBI was also significantly influenced by shelter type (X22 = 145.92, P < 0.0001). Females collected in animal sheds were significantly less likely to have fed on human hosts (22.3 ± 4%) than females collected in unoccupied houses (40.9 ± 6.8%; Chi-square post-hoc test: X21 = 21.6, P < 0.0001) or in human dwellings (74.5 ± 2%; Chi-square post-hoc test: X21 = 385, P < 0.0001). Females collected in human dwellings were also significantly more likely to have fed on human hosts than females collected in unoccupied houses (Chi-square post-hoc test: X21 = 96, P < 0.0001). HBI was significantly affected by the village (X22 = 139.5, P < 0.0001). However, in Soumousso only human dwellings were sampled confounding the effect of village and shelter type in this case. Therefore, we carried out an analysis on the human dwellings only to compare HBIs in the three villages. Mosquitoes were significantly less anthropophagic in Samendeni (56.5± 4%), compared to Soumousso (83.5±2.2%; Chi-square test: X21 =138.8, P < 0.0001) and Klesso (77.3±9 %; Chi - square test: X21 = 12.7, P = 0.0004). HBIs in Soumousso and Klesso were not significantly different (83.5±2.2% vs. 77.3±9 % respectively; Chi-square test: X21 = 1.8, P = 0.18).

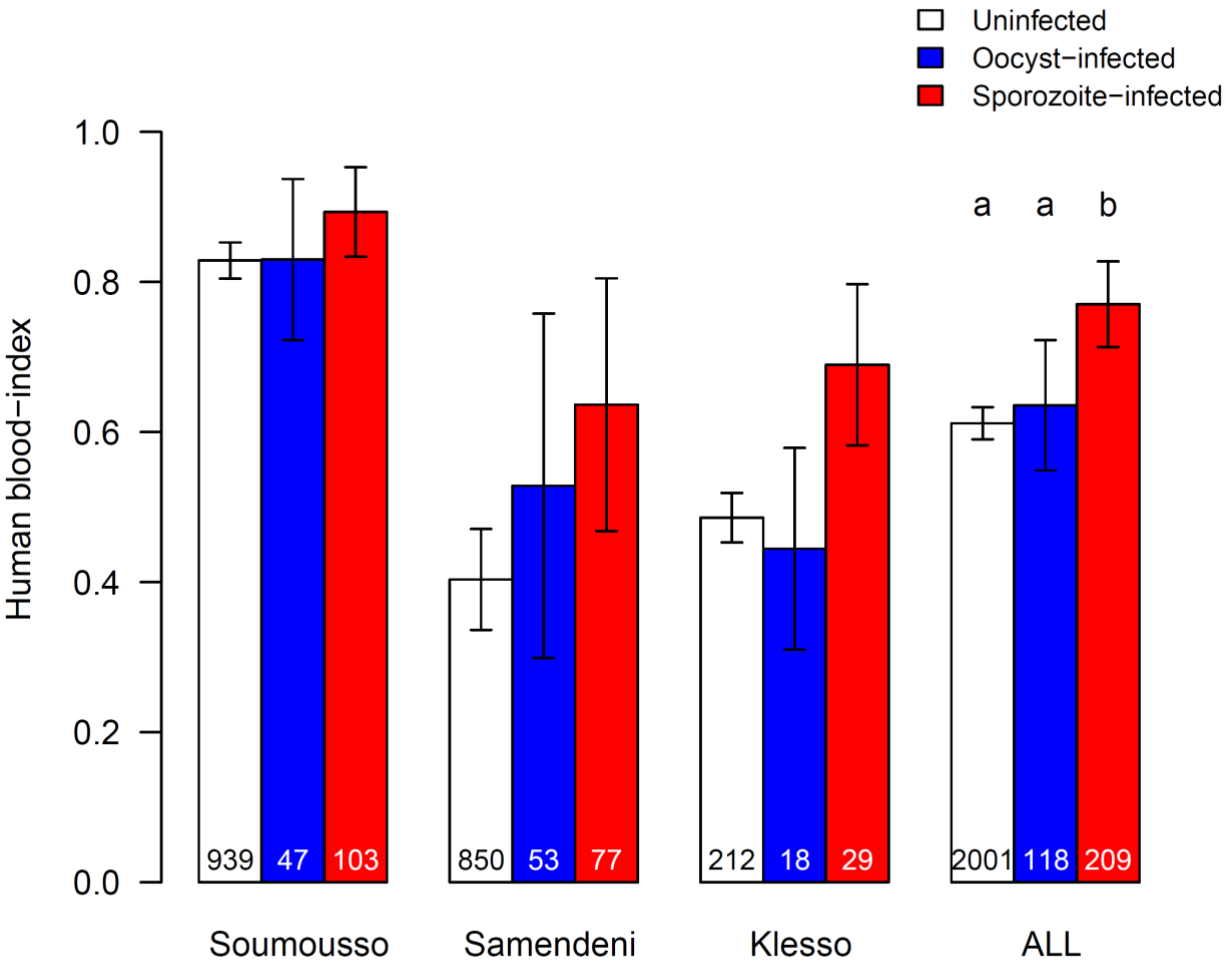

Figure 2. Effect of infection status on the human-blood index of *Anopheles gambiae s. l*. females expressed as the number of females fed on humans out of the total number of blood-fed females for the three sampled villages. Data show proportion ± 95% confidence intervals. Numbers in bars indicate the total numbers of mosquitoes. Different letters indicate differences between infection status (Chi-square post-hoc tests: sporozoite-infected *vs*. oocyst-infected females X^2_1=6.1, P=0.013; sporozoite-infected *vs*. uninfected females X^2_1=19.4, P<0.0001; oocyst-infected *vs*. uninfected females X^2_1=0.18, P= 0.67).

A significant species variation in HBI was observed (X^2_2 = 10.2, P = 0.006; Figure 3) with *Anopheles arabiensis* being significantly less anthropophagic (22.2 ± 15%, N=27, OR=3.5, 95% CI = 2.2-5.56, P = 0.007) than *An. gambiae* (54.8 ± 7.1%; N=186, OR=0.82, 95% CI = 0.71-0.95, P = 0.19) and *An. coluzzii* (55.1 ± 6.8%; N=205, OR=0.81, 95% CI = 0.71-0.94, P=0.14). Although HBI varied among mosquito species, sporozoite-infected individuals displayed the highest anthropophagy regardless of the species considered (infection status: species interaction: X^2_4 = 4, P = 0.42; Figure 3 and supplementary material).

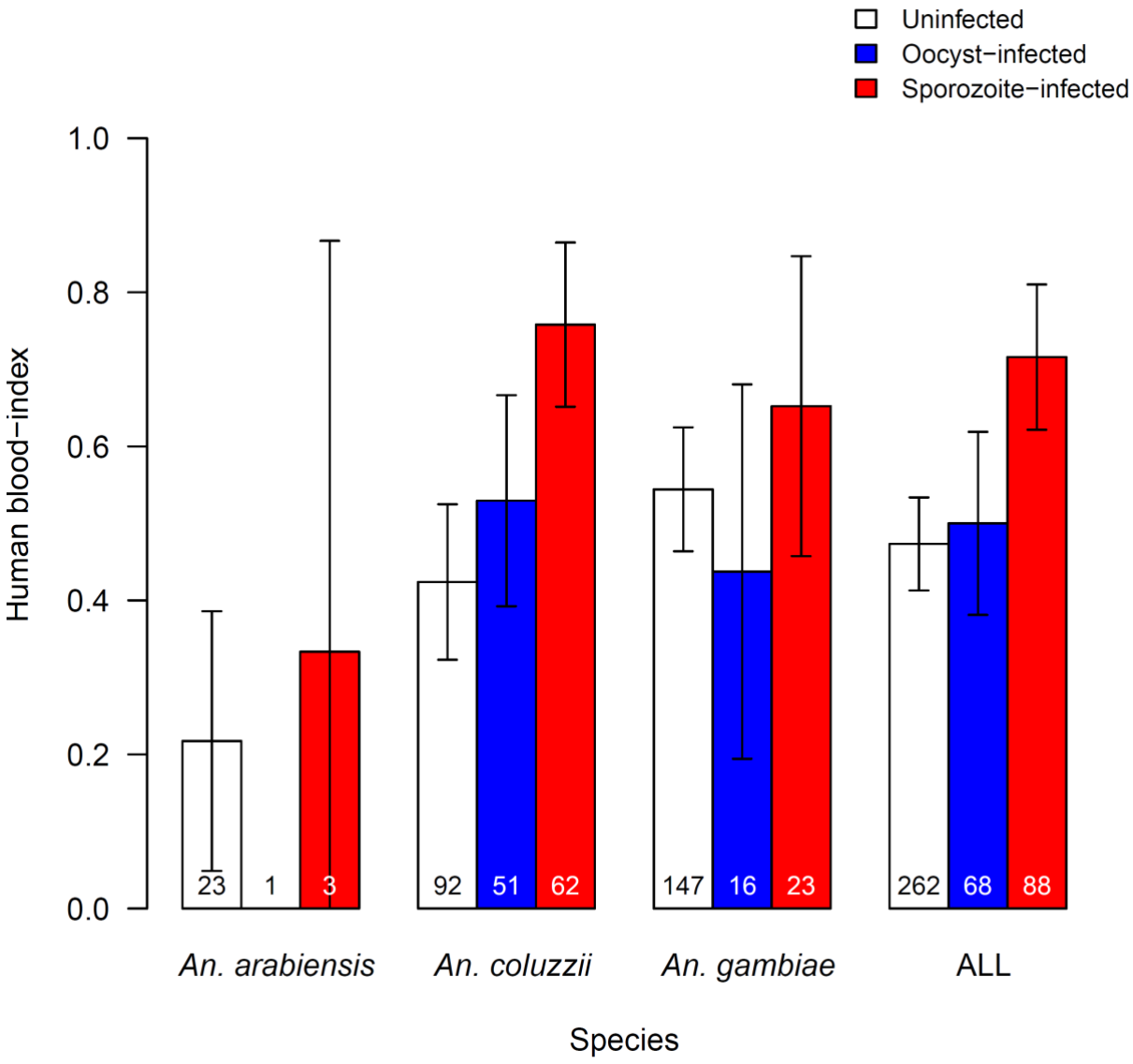

Figure 3. Effect of infection status and mosquito species on the human-blood index expressed as the proportion of females fed on humans or humans and animals out of the total of blood-fed females. Data show proportion ± 95% confidence intervals. Numbers in bars indicate the total numbers of mosquitoes.

Finally, HBI was not significantly affected by parity, a proxy used to estimate mosquito age (nulliparous females: 49.53 ± 9%, parous females: 45.6 ± 7.5%; $X^2_1 = 0.4$, P = 0.52).

Epidemiological consequences

To investigate the epidemiological impact of a higher HBI in infectious females compared to oocyst-infected and uninfected females, we built a mathematical model based on the experimental values observed in this study. This model assessed the impact of different HBIs on the Entomological Inoculation Rate (EIR, number of infectious bites received by a person during one year) at different mosquito lifespans and densities. In order to consider the heterogeneity of HBI values on epidemiological consequences, the HBI of susceptible mosquitoes was based on the average value whereas the HBI of exposed mosquitoes were assumed to be uniformly distributed within the confidence intervals of the HBI of oocyst-infected mosquitoes that were experimentally measured in this study. Then, the impact of HBI variation in infectious (sporozoite-infected) mosquitoes on parasite transmission potential was explored fully (Figure 4). For an

average mosquito lifespan of 15 days (Figure 4a), an HBI of 0.62 in infectious mosquitoes (similar to that of susceptible mosquitoes) resulted in an EIR of 4 at a low ratio of 1 (1 mosquito per human), while an HBI of 0.77 (as observed here in infectious mosquitoes) resulted in an EIR of 14. In other words, a 24% increase in HBI resulted in a 250% increase in EIR, everything else being equal. Transmission consequences were even larger when the human-to-mosquito ratios were higher (EIR = 5 *vs*. EIR = 19 with a ratio of 10 or 100, i.e. a 280% increase in EIR) but the size of the increase in EIR for sporozoite-infected mosquitoes declined with increasing mosquito longevity (Figure 4c, 4d, and supplementary material).

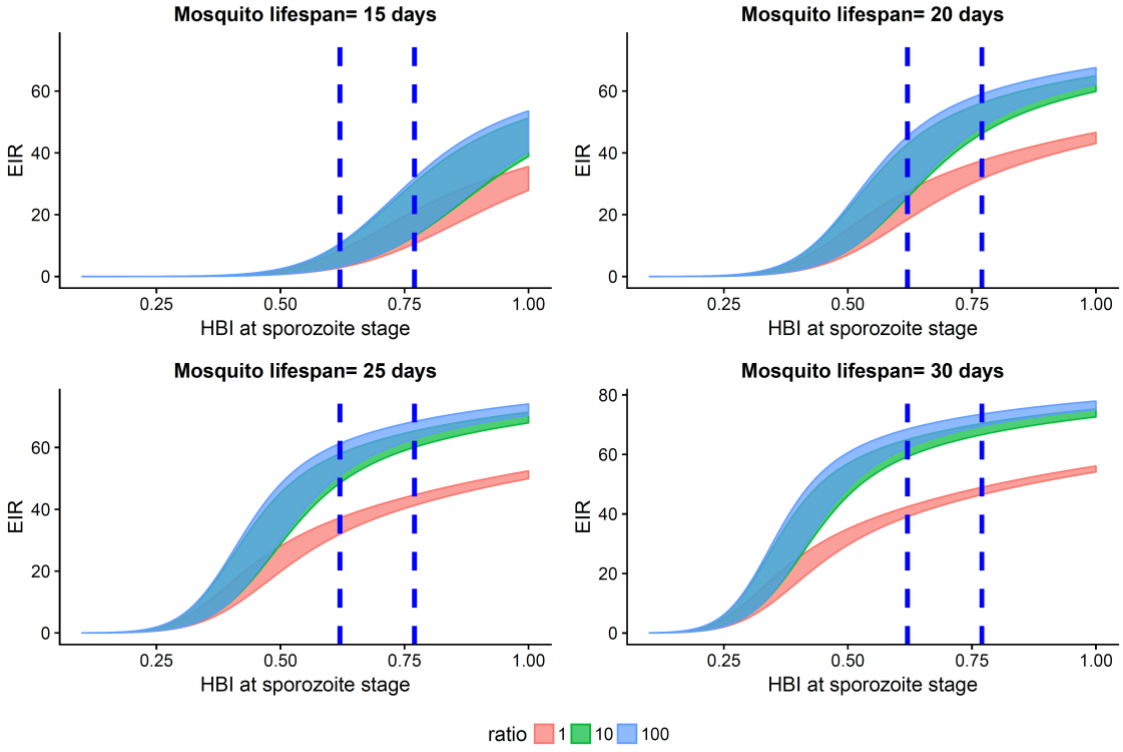

Figure 4. Expected epidemiological consequences of HBI variation for different values of mosquito lifespan and mosquito/human ratio. The X axis represents the range of values considered for the HBI of infectious (sporozoite-infected) mosquitoes and the Y axis is the Entomological Inoculation Rate (EIR, number of infectious bites received by a person over one year) when the HBI of exposed mosquitoes corresponds to the confidence intervals of the HBI of oocyst-infected mosquitoes that were experimentally measured in this study. The ribbons represent the possible EIR values for different HBI of sporozoite-infected mosquitoes according to the confidence interval of HBI in oocyst-infected mosquitoes (63.6% ± 5.7%) and for different values of the mosquito to human ratio. The dashed lines represents the average value measured for susceptible mosquitoes (0.62) and for sporozoite-infected mosquitoes (0.77). Ratio=adult mosquito/human densities.

Discussion

The mosquito host preference assays (experiment 1 using OBETs and BNTs,) showed that infected mosquitoes displayed similar long-range attraction toward human odour as uninfected individuals regardless of parasite developmental stages (oocyst *vs*. sporozoite), confirming previous laboratory results (Nguyen et al., 2017). However, consistent with the hypothesis of specific manipulation, the patterns of mosquito host selection (experiment 2 based on identification of mosquito blood-meal sources) showed that sporozoite-infected *An. coluzzi*, *An. gambiae* and *An. arabiensis* females were more likely to have fed

on human than oocyst-infected and uninfected individuals. By distinguishing sporozoite and oocyst infection, we ruled out the potential confounding effect of a mere intrinsic mosquito characteristic. Infected mosquitoes may indeed exhibit increased anthropophagy not because of being infected but just because of an innate preference for humans, thus making these mosquito individuals infected. Here, individuals infected with sporozoites displayed different HBI than individuals infected with oocysts, thus ruling out this possibility. Because *Plasmodium falciparum* takes about 10 to 18 days to complete its development (depending on temperature, (Nikolaev, 1935; Ohm et al., 2018; Shapiro et al., 2017) there is an increased likelihood of sporozoite infection as mosquitoes become older. This means that mosquito age could be a confounding factor of infection, with infected mosquitoes displaying increased HBI not because they harbour sporozoites but because they are older. Such an age effect could be mediated by specific physiological requirements in old mosquitoes or by a positive reinforcement (learning / memory) of feeding on humans. Our data does not support an age effect as we did not find a significant effect of parity (a proxy for age) on HBI (i.e. parous and nulliparous mosquito females displayed similar anthrophagy).

The precise mechanisms responsible for increased anthropophagy in sporozoite-infected mosquitoes is not yet clear, but at least three hypotheses can be proposed. First, malaria parasites might manipulate mosquito short-range behaviours only, whereas at longer range when mosquitoes rely mainly on CO_2 and other volatile odours (Cardé & Gibson, 2010; Gibson & Torr, 1999; Gillies, 1980; Mboera & Takken, 1997), sporozoite-infected mosquitoes display similar preferences to uninfected and oocyst-infected individuals. At short range, mosquitoes rely on other cues including visual stimuli, moisture, heat and skin emanations (Cardé & Gibson, 2010; Gibson & Torr, 1999; Takken & Verhulst, 2013). These stimuli can be host specific, and inform of host suitability for parasite development before the mosquito engages in selection and eventually in feeding. In addition to a possible preferential short-range attraction of sporozoite-infected mosquitoes toward host species suitable for parasite development, there could also be short-range repellence by unsuitable host species.

Second, the parasite may induce changes in the vector such as an alteration of microhabitat choice to spatially match the habitat of the suitable host. This could be achieved through parasite manipulation of mosquito endophagic/philic behaviours resulting in a higher degree of indoor -feeding and -resting of sporozoite-infected females. For example, infectious mosquitoes may exhibit an enhanced tendency to enter (or a decreased tendency to exit) house interstices regardless of emitted odours.

Third, the parasite may induce changes in the vector such as an alteration of time activity in order to temporally match the time of rest or activity of the suitable host. Mosquitoes exhibit circadian rhythms in many activities such as flight, host-seeking, swarming, egg-laying, etc. (Rund et al., 2016). There is mounting evidence that, following bed-net introduction, malaria vectors can display an increased tendency to feed outdoors (Russell et al., 2011) or bite earlier in the evening or later in the morning (Moiroux et al., 2012). Accordingly, *P. falciparum* could manipulate mosquito host-seeking rhythms in a way that increases bites on unprotected people. Testing this hypothesis would require sampling mosquitoes at distinct periods and comparing the proportion of uninfected, oocyst-infected and sporozoite-infected vectors among samples.

Sporozoite-induced change in mosquito host selection occurred in three major and related mosquito vectors, namely *An. coluzzii*, *An. gambiae* and *An. arabiensis*. This suggests that manipulation likely already occurred in the common ancestor of these three species and that the parasites might exploit a physiological pathway common to all three mosquito species to modify its vector host choice.

Transmission models generally assume that uninfected and infected vectors have similar preferences for human (D. Smith & Ellis McKenzie, 2004; D. L. Smith et al., 2012). This study suggests that this assumption may not be valid and that these models possibly underestimate transmission intensity. Our modelling approach confirms that HBI increases in infectious mosquitoes can have a dramatic impact on disease transmission. In particular, if we consider mosquito lifespans relevant to natural settings (i.e. 15 to 20 days; Charlwood et al., 1997; Gillies, 1961; Gillies & Wilkes, 1965; Killeen et al., 2000; Saul et al., 1990), the transmission potential was almost multiplied by 3 when the HBI increased from 0.62 to 0.77 *i.e*. the value observed for the infectious mosquitoes in this study. For many mosquito–*Plasmodium* associations including *An. gambiae s.l.-P. falciparum*, the duration of the parasite's development within the mosquito is as long as the insect vector's average lifespan (Charlwood et al., 1997; Gillies, 1961; Gillies & Wilkes, 1965; Killeen et al., 2000; Saul et al., 1990; World, 2014). This means that most mosquitoes do not live long enough to transmit the disease, and hence that feeds taken by infectious mosquitoes on unsuitable host

species would have disastrous consequences for parasite fitness. The model suggests that the benefits of specific manipulation should be particularly high in vectorial systems in which transmission opportunities are rare (short vector lifespan, relatively long parasite development period, and diverse blood sources).

In conclusion, our results suggest that the human malaria parasite *P. falciparum* evolved the ability to enhance transmission toward humans, the appropriate host species, by increasing mosquito anthropophagy (or decreasing zoophagy) with potentially profound public health consequences. Future laboratory and field studies will be essential to confirm these results and to better understand the epidemiological, ecological and evolutionary consequences of parasite manipulation of vector behaviours.

Acknowledgements

We would like to thank all volunteers for participating in this study as well as the local authorities for their support. We are very grateful to the IRSS staff in Burkina Faso for technical assistance. We thank Priscille Barreaux for discussion and comments. We are grateful to Alison Duncan, Ricardo S. Ramiro, Olivier Restif and an anonymous reviewer for comments and corrections which greatly improved the manuscript.

A previous version of this article has been reviewed and recommended by Peer Community In Evolutionary Biology (https://doi.org/10.24072/pci.evolbiol.100057).

Data availability

Raw data are available on zenodo: https://doi.org/10.5281/zenodo.1296744. Statistical analyses are available as supplementary information.

Competing interests

We have no competing interests.

Funding

This study was supported by the ANR grant no. 11-PDOC-006-01 and the European Community's Seventh Framework Program (FP7/2007–2013) under grant agreements no. 242095 and no.223736. BR is supported by the ANR project PANIC.

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Supplementary Material

Supplementary Results

Species subset

HBI was significantly affected by mosquito infection status (X^2_2 = 8.5, P = 0.014) with sporozoite-infected females being significantly more anthropophagic (71.6 ± 9.4%) than oocyst-infected females (50 ± 11.9%; Chi-square post-hoc tests: X^2_1=6.7, P = 0.0096) and uninfected females (47.3 ± 6%; Chi-square post-hoc tests: X^2_1 = 14.6, P =0.0001). There was no significant differences between oocyst-infected and uninfected females (Chi-square post-hoc test: X^2_1 =0.07, P = 0.8). HBI was significantly affected by the shelter type (X^2_2 = 50.8, P < 0.0001). In particular, the HBI in human dwelling females (73.7 ± 6.3%) was significantly higher than the HBI in unoccupied houses (34.7 ± 9.4%; Chi-square post-hoc tests: X^2_1=39, P < 0.0001) and animal sheds (37.3 ± 8.2%; X^2_1 = 40.9, P < 0.0001). The HBIs of unoccupied houses and animal sheds were not significantly different (Chi-square post-hoc test: X^2_1=0.07, P = 0.8). There was no significant interactions (infection status*shelter type: X^2_4 = 2.4, P = 0.66; shelter types*species: X^2_4 = 2.3, P = 0.67; three-way interaction: X^2_5 = 8, P = 0.15).

Epidemiological consequences

The impact of a larger HBI in infectious mosquitoes decreased with longer mosquito lifespan (20 days): the EIR increased by 54% at low mosquito density (EIR = 22 *vs.* EIR = 34), which was similar at larger densities, 51% increase in both cases (EIR = 33 *vs*. EIR = 50 for a ratio mosquito/human at 10 and EIR = 35 vs EIR = 53 for a ratio at 100). The pattern is similar with a mosquito lifespan of 25 days: an increase of 23% at a ratio of 1 (EIR =34 *vs*. EIR =42) and increases of 16% at a ratio of 10 (EIR =53 *vs.* EIR =62) or a ratio of 100 (EIR =56 *vs*. EIR =65) . Pattern which is kept constant as well with a mosquito lifespan of 30 days: a 14% increase with a ratio of 1 (EIR = 41 *vs*. EIR =47), a 7% increase with a ratio of 10 (EIR =63 *vs*. EIR =68), and a 5% increase with a ratio 100 (EIR =67 *vs*. EIR =71).

Supplementary Table

Table S1: Parameters used in the mathematical model.

Parameter	Unit	Value
a (biting frequency)	days.ind-1	4
b (mosquito probability to get infected)	%	0.5
εs (human biting rate of susceptible mosquitoes)	%	0.62
γ (extrinsic incubation period)	days.ind-1	14
μ (mosquito population dying rate)	days.ind-1	variable
c (human probability to develop infection)	%	0.5
εi (human biting rate of infectious mosquitoes)	%	variable

1/δ (human infectious period)	day	30

Supplementary figures

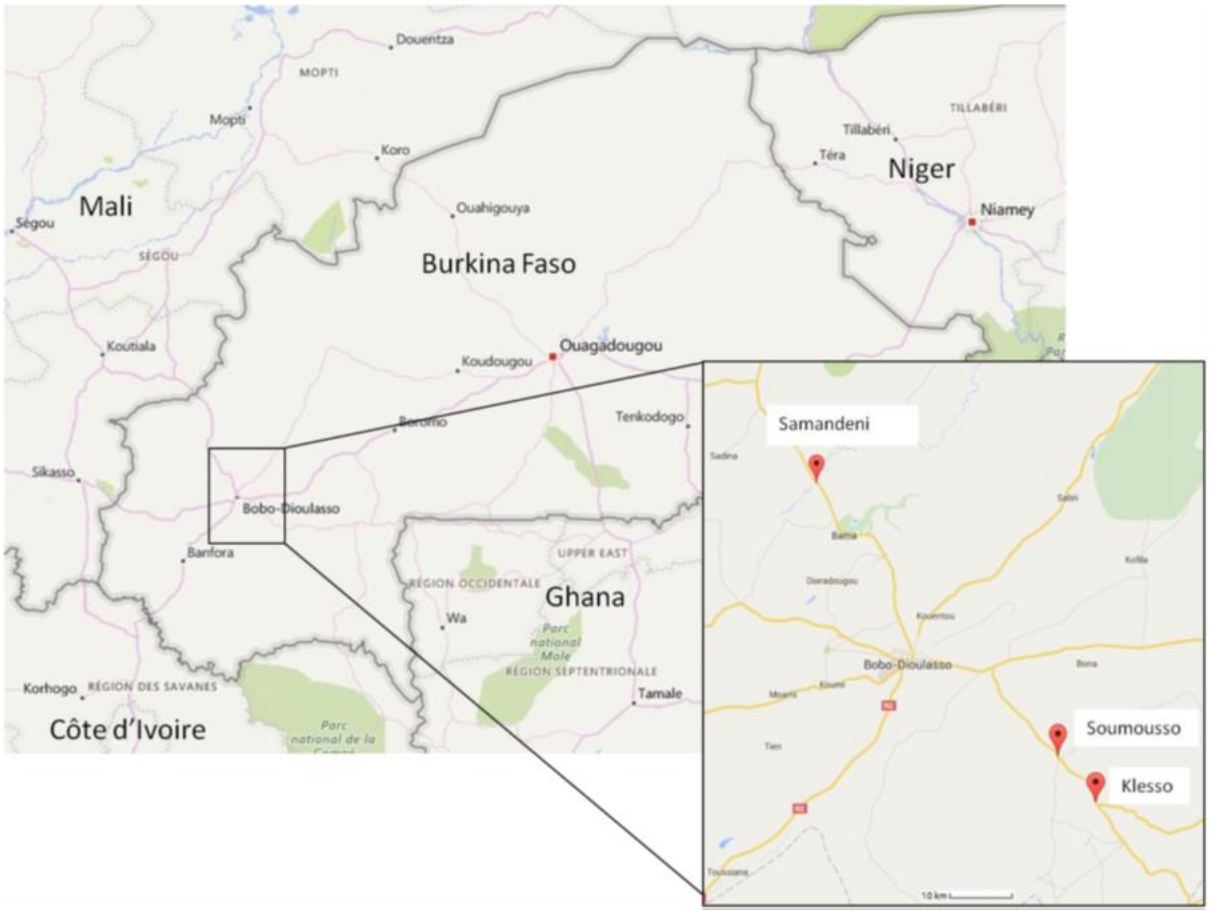

Figure S1. Study collection sites: Soumousso (11°23’14”N, 4°24’42”W), Klesso (10°56’40.5”N, 3°59’09.9”W), Samendeni (11°27’14.3”N, 4°27’37.6”W)

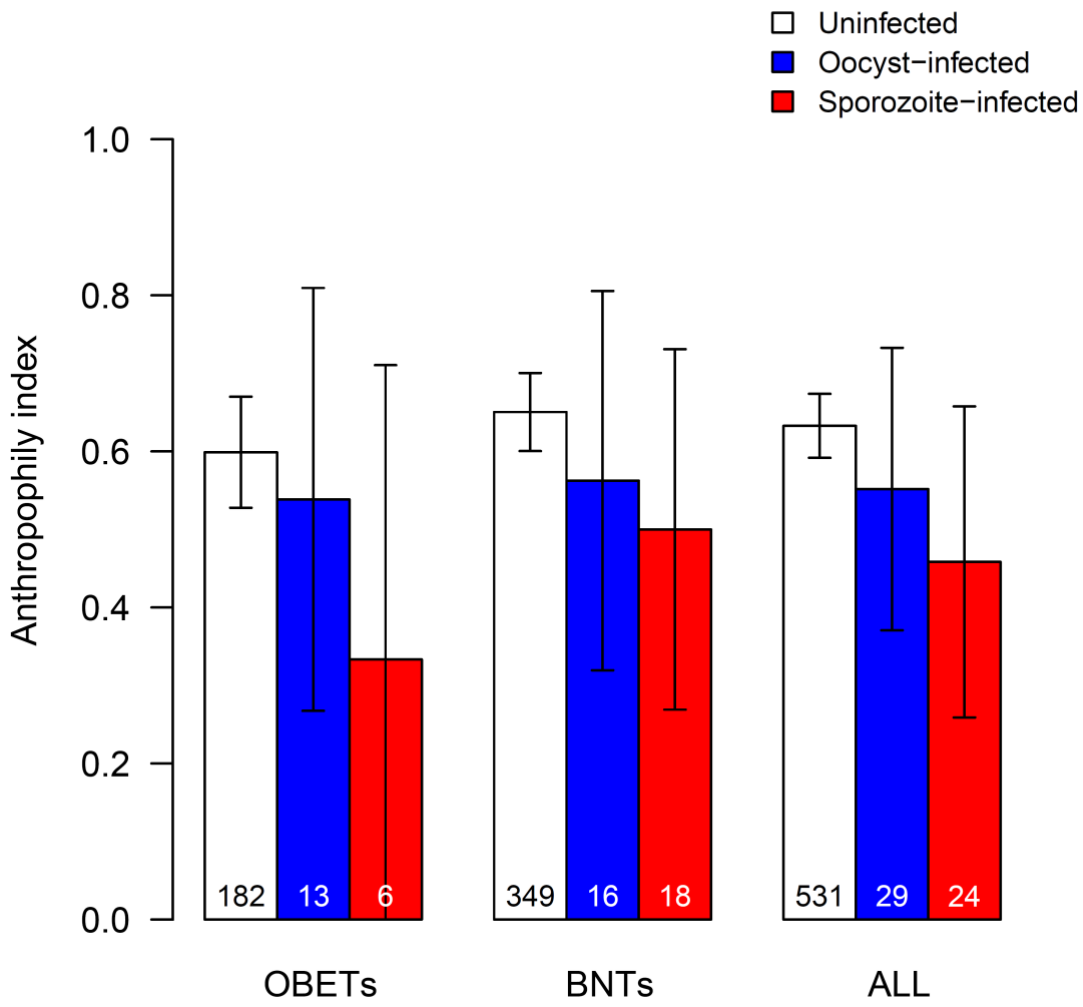

Figure S2. Effect of infection status on the anthropophily index of *Anopheles gambiae s. l*. females expressed as the proportion of females caught in the human-baited traps out of the total number retrieved from both human- and calf- baited traps. Data show proportion ± 95% confidence interval. Numbers in bars indicate the total numbers of mosquitoes in both traps.

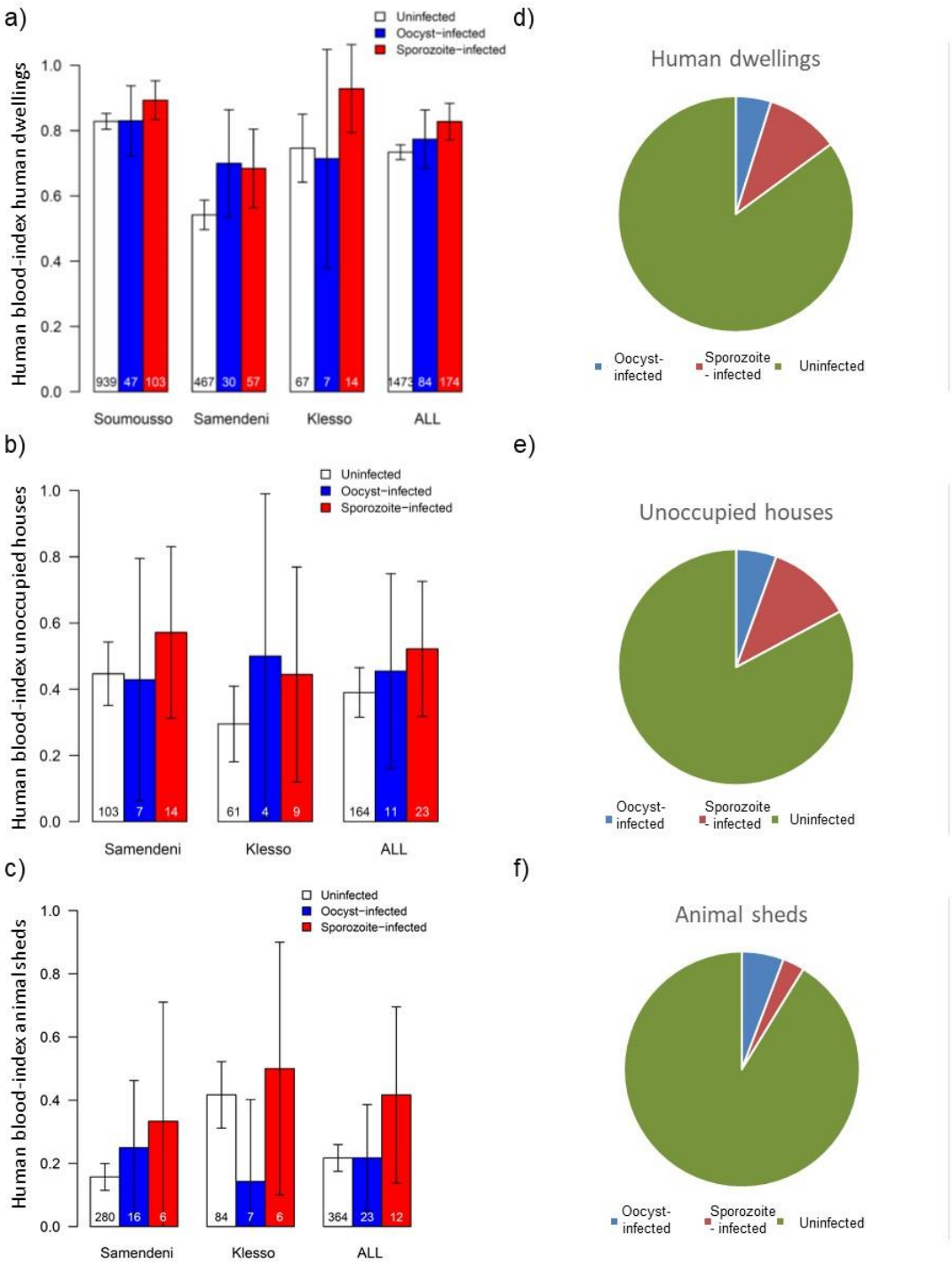

Figure S3. Effect of infection status on the human-blood index of *Anopheles gambiae s. l*. females expressed as the number of females fed on humans or human-animal mixed blood meals over the total number of blood-fed females in the different village samples in a) human dwellings, b) unoccupied houses and c) animal sheds. Data show proportion ± 95% confidence intervals. Numbers in bars indicate the total numbers of mosquitoes in both traps. Relative proportions of females according to their infection status in d) human dwellings, e) unoccupied houses and f) animal sheds.

Supplementary material – Statistical analyses

```
########## Analyses Experiment 1: Mosquito host preference #######

############ GLM mosquito infection by collection method ############
```

```
>  m1=glm(choice2~infection*collection,family=binomial)
> summary(m1)

Call:
glm(formula = choice2 ~ infection * collection, family = binomial)

Deviance Residuals:
    Min       1Q   Median       3Q      Max
-1.4823  -0.9275  -0.9275   1.3517   1.4499

Coefficients:
                                   Estimate Std. Error z value Pr(>|z|)
(Intercept)                        -0.25131    0.50395  -0.499    0.618
infectionspz                        0.25131    0.69007   0.364    0.716
infectionuninfected                -0.36961    0.51630  -0.716    0.474
collectiontente                     0.09716    0.75066   0.129    0.897
infectionspz:collectiontente        0.59598    1.23924   0.481    0.631
infectionuninfected:collectiontente 0.12288    0.77393   0.159    0.874

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 774.72  on 583  degrees of freedom
Residual deviance: 769.34  on 578  degrees of freedom
AIC: 781.34

Number of Fisher Scoring iterations: 4

>  m2=glm(choice2~infection+collection,family=binomial)
> anova(m1,m2,test="Chi")
Analysis of Deviance Table

Model 1: choice2 ~ infection * collection
Model 2: choice2 ~ infection + collection
  Resid. Df Resid. Dev Df Deviance Pr(>Chi)
1       578     769.34
2       580     769.60 -2 -0.26068   0.8778
> summary(m2)

Call:
glm(formula = choice2 ~ infection + collection, family = binomial)

Deviance Residuals:
    Min       1Q   Median       3Q      Max
-1.3246  -0.9263  -0.9263   1.3491   1.4513

Coefficients:
                   Estimate Std. Error z value Pr(>|z|)
(Intercept)         -0.3111     0.3828  -0.813    0.416
infectionspz         0.4213     0.5563   0.757    0.449
infectionuninfected -0.3131     0.3851  -0.813    0.416
collectiontente      0.2293     0.1794   1.278    0.201

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 774.72  on 583  degrees of freedom
Residual deviance: 769.60  on 580  degrees of freedom
AIC: 777.6

Number of Fisher Scoring iterations: 4

>  m3=glm(choice2~collection,family=binomial)
> anova(m3,m2,test="Chi")
Analysis of Deviance Table

Model 1: choice2 ~ collection
Model 2: choice2 ~ infection + collection
  Resid. Df Resid. Dev Df Deviance Pr(>Chi)
1       582     773.18
2       580     769.60  2   3.5794    0.167
>  m4=glm(choice2~1,family=binomial)
> anova(m3,m4,test="Chi")
Analysis of Deviance Table

Model 1: choice2 ~ collection
Model 2: choice2 ~ 1
  Resid. Df Resid. Dev Df Deviance Pr(>Chi)
```

```
1       582     773.18
2       583     774.72 -1  -1.5442    0.214
>


########## models with intercepts in each category of individuals ############



>   t=read.table("terrainangtraps.txt",header=T)
>   attach(t)
> tooc=subset(t,infection=="oocyst")
> detach(t)
> attach(tooc)
> m=glm(choice2~1,family=binomial)
> summary(m)

Call:
glm(formula = choice2 ~ 1, family = binomial)

Deviance Residuals:
   Min      1Q  Median      3Q     Max
-1.091  -1.091  -1.091   1.267   1.267

Coefficients:
            Estimate Std. Error z value Pr(>|z|)
(Intercept)  -0.2076     0.3734  -0.556    0.578

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 39.892  on 28  degrees of freedom
Residual deviance: 39.892  on 28  degrees of freedom
AIC: 41.892

Number of Fisher Scoring iterations: 3

> OR=exp(-0.2076)
> OR
[1] 0.812532
> CI1=exp(-0.2076-0.3734)
> CI1
[1] 0.5593387
> CI2=exp(-0.2076+0.3734)
> CI2
[1] 1.180337


> detach(tooc)
> attach(t)
> tspz=subset(t,infection=="spz")
> detach(t)
> attach(tspz)

> m=glm(choice2~1,family=binomial)
> summary(m)

Call:
glm(formula = choice2 ~ 1, family = binomial)

Deviance Residuals:
   Min      1Q  Median      3Q     Max
-1.249  -1.249   1.107   1.107   1.107

Coefficients:
            Estimate Std. Error z value Pr(>|z|)
(Intercept)   0.1671     0.4097   0.408    0.683

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 33.104  on 23  degrees of freedom
Residual deviance: 33.104  on 23  degrees of freedom
AIC: 35.104

Number of Fisher Scoring iterations: 3

> OR=exp(0.1671)
> OR
```

```
[1] 1.181872
> CI1=exp(0.1671+0.4097)
> CI1
[1] 1.780332
> CI2=exp(0.1671-0.4097)
> CI2
[1] 0.7845853




> detach(tspz)
> attach(t)
> tun=subset(t,infection=="uninfected")
> detach(t)
> attach(tun)

> m=glm(choice2~1,family=binomial)
> summary(m)

Call:
glm(formula = choice2 ~ 1, family = binomial)

Deviance Residuals:
    Min       1Q   Median       3Q      Max
-0.9567  -0.9567  -0.9567   1.4155   1.4155

Coefficients:
            Estimate Std. Error z value Pr(>|z|)
(Intercept) -0.54411    0.09002  -6.044  1.5e-09 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 698.23  on 530  degrees of freedom
Residual deviance: 698.23  on 530  degrees of freedom
AIC: 700.23

Number of Fisher Scoring iterations: 4

> OR=exp(-0.54411)
> OR
[1] 0.5803581
> CI1=exp(-0.54411+0.09002)
> CI1
[1] 0.6350256
> CI2=exp(-0.54411-0.09002)
> CI2
[1] 0.5303967
>




####### Analyses Experiment 2: Mosquito blood-feeding pattern #######

> m1=glm(choice2~infection*origin+infection*village,family=binomial)
> summary(m1)

Call:
glm(formula = choice2 ~ infection * origin + infection * village,
    family = binomial)

Deviance Residuals:
    Min       1Q   Median       3Q      Max
-1.8258  -0.6133  -0.6133   0.8630   2.1151

Coefficients:
                                  Estimate Std. Error z value Pr(>|z|)
(Intercept)                        1.34947    0.66654   2.025 0.042908 *
infectionspz                      -1.31907    0.92452  -1.427 0.153650
infectionuninfected               -0.55363    0.68931  -0.803 0.421878
originMH                          -2.13075    0.62341  -3.418 0.000631 ***
originMI                          -1.10487    0.79012  -1.398 0.162006
villagesamandeni                  -0.09770    0.61400  -0.159 0.873574
villagesoumousso                  -0.80284    0.72406  -1.109 0.267516
```

```
infectionspz:originMH                 0.57376    0.92201   0.622 0.533749
infectionuninfected:originMH          0.48010    0.64327   0.746 0.455463
infectionspz:originMI                 0.60162    1.07679   0.559 0.576354
infectionuninfected:originMI          0.34986    0.81689   0.428 0.668446
infectionspz:villagesamandeni         0.72644    0.80107   0.907 0.364497
infectionuninfected:villagesamandeni  0.75924    0.63728   1.191 0.233506
infectionspz:villagesoumousso         0.20553    0.94357   0.218 0.827565
infectionuninfected:villagesoumousso  0.08233    0.75010   0.110 0.912597
---
Signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 3075.1  on 2327  degrees of freedom
Residual deviance: 2502.3  on 2313  degrees of freedom
AIC: 2532.3

Number of Fisher Scoring iterations: 4

> m2=glm(choice2~origin+infection*village,family=binomial)
> anova(m1,m2,test="Chi")
Analysis of Deviance Table

Model 1: choice2 ~ infection * origin + infection * village
Model 2: choice2 ~ origin + infection * village
  Resid. Df Resid. Dev Df Deviance Pr(>Chi)
1      2313     2502.3
2      2317     2503.0 -4 -0.72688    0.948
> summary(m2)

Call:
glm(formula = choice2 ~ origin + infection * village, family = binomial)

Deviance Residuals:
    Min       1Q   Median       3Q      Max
-1.8336  -0.6133  -0.6133   0.8600   2.1151

Coefficients:
                                    Estimate Std. Error z value Pr(>|z|)
(Intercept)                           1.0722     0.5150   2.082   0.0373 *
originMH                             -1.6781     0.1498 -11.204  < 2e-16 ***
originMI                             -0.7556     0.1921  -3.933 8.41e-05 ***
infectionspz                         -0.9040     0.6589  -1.372   0.1701
infectionuninfected                  -0.2680     0.5269  -0.509   0.6110
villagesamandeni                     -0.1348     0.5860  -0.230   0.8181
villagesoumousso                     -0.9782     0.6424  -1.523   0.1278
infectionspz:villagesamandeni         0.7649     0.7624   1.003   0.3158
infectionuninfected:villagesamandeni  0.8056     0.6085   1.324   0.1855
infectionspz:villagesoumousso         0.3642     0.8286   0.440   0.6602
infectionuninfected:villagesoumousso  0.2768     0.6601   0.419   0.6750
---
Signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 3075.1  on 2327  degrees of freedom
Residual deviance: 2503.0  on 2317  degrees of freedom
AIC: 2525

Number of Fisher Scoring iterations: 4

> m3=glm(choice2~origin+infection+village,family=binomial)
> anova(m3,m2,test="Chi")
Analysis of Deviance Table

Model 1: choice2 ~ origin + infection + village
Model 2: choice2 ~ origin + infection * village
  Resid. Df Resid. Dev Df Deviance Pr(>Chi)
1      2321     2505.3
2      2317     2503.0  4     2.29   0.6826
> summary(m3)

Call:
glm(formula = choice2 ~ origin + infection + village, family = binomial)

Deviance Residuals:
    Min       1Q   Median       3Q      Max
```

```
-1.8244  -0.6175  -0.6175   0.8471   2.1436

Coefficients:
                    Estimate Std. Error z value Pr(>|z|)
(Intercept)           0.5994     0.2629   2.280   0.0226 *
originMH             -1.6762     0.1494 -11.221  < 2e-16 ***
originMI             -0.7611     0.1916  -3.972 7.12e-05 ***
infectionspz         -0.3902     0.2809  -1.389   0.1648
infectionuninfected   0.2408     0.2235   1.077   0.2814
villagesamandeni      0.6142     0.1562   3.933 8.40e-05 ***
villagesoumousso     -0.7245     0.1801  -4.022 5.78e-05 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 3075.1  on 2327  degrees of freedom
Residual deviance: 2505.3  on 2321  degrees of freedom
AIC: 2519.3

Number of Fisher Scoring iterations: 4

> m4=glm(choice2~origin+village,family=binomial)
> anova(m3,m4,test="Chi")
Analysis of Deviance Table

Model 1: choice2 ~ origin + infection + village
Model 2: choice2 ~ origin + village
  Resid. Df Resid. Dev Df Deviance Pr(>Chi)
1      2321     2505.3
2      2323     2518.3 -2  -13.007 0.001498 **
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
> m5=glm(choice2~infection+village,family=binomial)
> anova(m3,m5,test="Chi")
Analysis of Deviance Table

Model 1: choice2 ~ origin + infection + village
Model 2: choice2 ~ infection + village
  Resid. Df Resid. Dev Df Deviance  Pr(>Chi)
1      2321     2505.3
2      2323     2651.2 -2  -145.92 < 2.2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
> m6=glm(choice2~origin+infection,family=binomial)
> anova(m3,m6,test="Chi")
Analysis of Deviance Table

Model 1: choice2 ~ origin + infection + village
Model 2: choice2 ~ origin + infection
  Resid. Df Resid. Dev Df Deviance  Pr(>Chi)
1      2321     2505.3
2      2323     2644.8 -2   -139.5 < 2.2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1



> table(choice2,village)
       village
choice2 klesso samandeni soumousso
  H        131       420       909
  other    128       560       180


> ######## klesso-samendeni
> x=matrix(c(131,128,420,560),ncol=2)
>  chisq.test(x)

        Pearson's Chi-squared test with Yates' continuity correction

data:  x
X-squared = 4.639, df = 1, p-value = 0.03125

> ######## klesso-soumousso
> x=matrix(c(131,128,909,180),ncol=2)
>  chisq.test(x)
```

```
        Pearson's Chi-squared test with Yates' continuity correction

data:  x
X-squared = 126.5548, df = 1, p-value < 2.2e-16

> ######## samendeni-soumousso
> x=matrix(c(420,560,909,180),ncol=2)
>  chisq.test(x)

        Pearson's Chi-squared test with Yates' continuity correction

data:  x
X-squared = 368.5804, df = 1, p-value < 2.2e-16

> table(choice2,origin)
       origin
choice2   CA   MH   MI
  H       89 1290   81
  other  310  441  117

> ######## CA-MH
> x=matrix(c(89,310,1290,441),ncol=2)
>  chisq.test(x)

        Pearson's Chi-squared test with Yates' continuity correction

data:  x
X-squared = 385.0446, df = 1, p-value < 2.2e-16

> ######## CA-MI
> x=matrix(c(89,310,81,117),ncol=2)
>  chisq.test(x)

        Pearson's Chi-squared test with Yates' continuity correction

data:  x
X-squared = 21.5821, df = 1, p-value = 3.39e-06

> ######## MH-MI
> x=matrix(c(1290,441,81,117),ncol=2)
>  chisq.test(x)

        Pearson's Chi-squared test with Yates' continuity correction

data:  x
X-squared = 96.0216, df = 1, p-value < 2.2e-16

>

> #### HBI ooc vs uninfected####

> x=matrix(c(75,43,1224,777),ncol=2)
> chisq.test(x)

        Pearson's Chi-squared test with Yates' continuity correction

data:  x
X-squared = 0.177, df = 1, p-value = 0.674



> #### HBI spz vs uninfected####

> x=matrix(c(161,48,1224,777),ncol=2)
> chisq.test(x)

        Pearson's Chi-squared test with Yates' continuity correction

data:  x
X-squared = 19.6844, df = 1, p-value = 9.134e-06



> #### HBI spz vs ooc####

> x=matrix(c(161,48,75,43),ncol=2)
```

```
> chisq.test(x)

        Pearson's Chi-squared test with Yates' continuity correction

data:  x
X-squared = 6.1632, df = 1, p-value = 0.01304

############### intercept model by infectious status


>  t=read.table("terrainangmaisons.txt",header=T)
>  attach(t)
>  summary(t)
> tooc=subset(t,infection=="oocyst")
> detach(t)
> attach(tooc)
> summary(tooc)

> m=glm(choice2~1,family=binomial)
> summary(m)

Call:
glm(formula = choice2 ~ 1, family = binomial)

Deviance Residuals:
   Min      1Q  Median      3Q     Max
-0.952  -0.952  -0.952   1.421   1.421

Coefficients:
            Estimate Std. Error z value Pr(>|z|)
(Intercept)  -0.5563     0.1913  -2.908  0.00364 **
---
Signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 154.8  on 117  degrees of freedom
Residual deviance: 154.8  on 117  degrees of freedom
AIC: 156.8

Number of Fisher Scoring iterations: 4

> OR=exp(-0.5563)
> OR
[1] 0.5733265
> CI1=exp(-0.5563+0.1913)
> CI1
[1] 0.6941967
> CI2=exp(-0.5563-0.1913)
> CI2
[1] 0.4735016

> detach(tooc)
> attach(t)
> tspz=subset(t,infection=="spz")
> detach(t)
> attach(tspz)
> summary(tspz)

> m=glm(choice2~1,family=binomial)
> summary(m)

Call:
glm(formula = choice2 ~ 1, family = binomial)

Deviance Residuals:
    Min       1Q   Median       3Q      Max
-0.7224  -0.7224  -0.7224  -0.7224   1.7153

Coefficients:
            Estimate Std. Error z value Pr(>|z|)
(Intercept)  -1.2102     0.1645  -7.359 1.85e-13 ***
---
Signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

(Dispersion parameter for binomial family taken to be 1)
```

```
    Null deviance: 225.25  on 208  degrees of freedom
Residual deviance: 225.25  on 208  degrees of freedom
AIC: 227.25

Number of Fisher Scoring iterations: 4

> OR=exp(-1.2102)
> OR
[1] 0.2981376
> CI1=exp(-1.2102+0.1645)
> CI1
[1] 0.3514457
> CI2=exp(-1.2102-0.1645)
> CI2
[1] 0.2529155

> detach(tspz)
> attach(t)
> tun=subset(t,infection=="uninfected")
> detach(t)
> attach(tun)
> summary(tun)

> m=glm(choice2~1,family=binomial)
> summary(m)

Call:
glm(formula = choice2 ~ 1, family = binomial)

Deviance Residuals:
    Min       1Q   Median       3Q      Max
-0.9915  -0.9915  -0.9915   1.3755   1.3755

Coefficients:
            Estimate Std. Error z value Pr(>|z|)
(Intercept) -0.45444    0.04587  -9.907   <2e-16 ***
---
Signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 2673.3  on 2000  degrees of freedom
Residual deviance: 2673.3  on 2000  degrees of freedom
AIC: 2675.3

Number of Fisher Scoring iterations: 4

> OR=exp(-0.45444)
> OR
[1] 0.6348034
> CI1=exp(-0.45444+0.04587)
> CI1
[1] 0.6645999
> CI2=exp(-0.45444-0.04587)
> CI2
[1] 0.6063427
>



#### Analyses subset of 150 randomly selected uninfected individuals####

install.packages("dplyr")
library(dplyr)
b                    =                          rbind                          (hbiam
[which(hbiam$infection=="oocyst"),],hbiam[which(hbiam$infection=="spz"),],sample_n(hbiam[whic
h(hbiam$infection=="uninfected"),],150))
summary(b)
attach(b)
mod1<-glm(choice2~infection,binomial)
anova(mod1,test="Chi")
library(multcomp)
glht.mod <- glht(mod1, mcp(infection = "Tukey"))
summary(glht.mod)

OUTPUT:
```

```
> anova(mod1,test="Chi")
Analysis of Deviance Table

Model: binomial, link: logit

Response: choice2

Terms added sequentially (first to last)


          Df Deviance Resid. Df Resid. Dev Pr(>Chi)
NULL                        476     590.83
infection  2   11.567       474     579.26 0.003077 **
---

> summary(glht.mod)

     Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts


Fit: glm(formula = choice2 ~ infection, family = binomial)

Linear Hypotheses:
                          Estimate Std. Error z value Pr(>|z|)
spz - oocyst == 0        -0.65392    0.25226  -2.592  0.02570 *
uninfected - oocyst == 0  0.06674    0.25473   0.262  0.96283
uninfected - spz == 0     0.72066    0.23525   3.063  0.00614 **
---
Signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1
(Adjusted p values reported -- single-step method)


######### Analyses of HBI in human dwellings only ##########

> m1=glm(choice2~infection*village,family=binomial)
> summary(m1)

Call:
glm(formula = choice2 ~ infection * village, family = binomial)

Deviance Residuals:
    Min       1Q   Median       3Q      Max
-1.1072  -0.6133  -0.6133   1.2493   2.2974

Coefficients:
                                    Estimate Std. Error z value Pr(>|z|)
(Intercept)                         -0.91629    0.83666  -1.095    0.273
infectionspz                        -1.64866    1.33204  -1.238    0.216
infectionuninfected                 -0.16252    0.88251  -0.184    0.854
villagesamandeni                     0.06899    0.92668   0.074    0.941
villagesoumousso                    -0.66783    0.92230  -0.724    0.469
infectionspz:villagesamandeni        1.72276    1.41925   1.214    0.225
infectionuninfected:villagesamandeni 0.84240    0.97272   0.866    0.386
infectionspz:villagesoumousso        1.10888    1.42364   0.779    0.436
infectionuninfected:villagesoumousso 0.17132    0.96797   0.177    0.860

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 1964.7  on 1730  degrees of freedom
Residual deviance: 1816.7  on 1722  degrees of freedom
AIC: 1834.7

Number of Fisher Scoring iterations: 4

> m2=glm(choice2~infection+village,family=binomial)
> anova(m1,m2,test="Chi")
Analysis of Deviance Table

Model 1: choice2 ~ infection * village
Model 2: choice2 ~ infection + village
  Resid. Df Resid. Dev Df Deviance Pr(>Chi)
1      1722     1816.7
2      1726     1819.3 -4  -2.6054   0.6259
> summary(m2)
```

```
Call:
glm(formula = choice2 ~ infection + village, family = binomial)

Deviance Residuals:
    Min       1Q   Median       3Q      Max
-1.1005  -0.6182  -0.6182   1.2563   2.1416

Coefficients:
                    Estimate Std. Error z value Pr(>|z|)
(Intercept)          -1.4252     0.3631  -3.925 8.69e-05 ***
infectionspz         -0.3177     0.3414  -0.931 0.352112
infectionuninfected   0.3112     0.2785   1.117 0.263790
villagesamandeni      0.9303     0.2696   3.451 0.000559 ***
villagesoumousso     -0.4440     0.2686  -1.653 0.098280 .
---
Signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 1964.7  on 1730  degrees of freedom
Residual deviance: 1819.3  on 1726  degrees of freedom
AIC: 1829.3

Number of Fisher Scoring iterations: 4

> m3=glm(choice2~village,family=binomial)
> anova(m3,m2,test="Chi")
Analysis of Deviance Table

Model 1: choice2 ~ village
Model 2: choice2 ~ infection + village
  Resid. Df Resid. Dev Df Deviance Pr(>Chi)
1      1728     1829.4
2      1726     1819.3  2   10.099 0.006411 **
---
Signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1
> m4=glm(choice2~infection,family=binomial)
> anova(m4,m2,test="Chi")
Analysis of Deviance Table

Model 1: choice2 ~ infection
Model 2: choice2 ~ infection + village
  Resid. Df Resid. Dev Df Deviance  Pr(>Chi)
1      1728     1956.6
2      1726     1819.3  2   137.27 < 2.2e-16 ***
---
Signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

> table(choice2,village)
       village
choice2 klesso samandeni soumousso
  H         68       313       909
  other     20       241       180


> ######## klesso-samendeni
> x=matrix(c(68,20,313,241),ncol=2)
>  chisq.test(x)

        Pearson's Chi-squared test with Yates' continuity correction

data:  x
X-squared = 12.7365, df = 1, p-value = 0.0003586


> ######## klesso-soumousso
> x=matrix(c(68,20,909,180),ncol=2)
>  chisq.test(x)

        Pearson's Chi-squared test with Yates' continuity correction

data:  x
X-squared = 1.8001, df = 1, p-value = 0.1797

> ######## samendeni-soumousso
> x=matrix(c(313;241,909,180),ncol=2)
Error: unexpected ';' in "x=matrix(c(313;"
```

```
> x=matrix(c(313,241,909,180),ncol=2)
>  chisq.test(x)

        Pearson's Chi-squared test with Yates' continuity correction

data:  x
X-squared = 138.7652, df = 1, p-value < 2.2e-16


> table(choice2,infection)
       infection
choice2 oocyst  spz uninfected
  H         65  144       1081
  other     19   30        392
> ######ooc-spz
> x=matrix(c(65,19,144,30),ncol=2)
>  chisq.test(x)

        Pearson's Chi-squared test with Yates' continuity correction

data:  x
X-squared = 0.744, df = 1, p-value = 0.3884

> ######ooc-uninf
> x=matrix(c(65,19,1081,392),ncol=2)
>  chisq.test(x)

        Pearson's Chi-squared test with Yates' continuity correction

data:  x
X-squared = 0.4629, df = 1, p-value = 0.4963

> ######spz-uninf
> x=matrix(c(144,30,1081,392),ncol=2)
>  chisq.test(x)

        Pearson's Chi-squared test with Yates' continuity correction

data:  x
X-squared = 6.6875, df = 1, p-value = 0.009709

>




########### Analyses by mosquito species ##########

> m1=glm(choice2~infection*species.mol*origin,family=binomial)
> summary(m1)

Call:
glm(formula = choice2 ~ infection * species.mol * origin, family = binomial)

Deviance Residuals:
    Min       1Q   Median       3Q      Max
-2.0963  -0.9854  -0.4366   0.8733   2.1899

Coefficients: (3 not defined because of singularities)
                                        Estimate Std. Error z value
(Intercept)                             15.56607 1455.39751   0.011
infectionspz                           -31.13214 2058.24292  -0.015
infectionuninfected                    -13.48663 1455.39790  -0.009
species.molc                           -14.55447 1455.39763  -0.010
species.molg                           -13.95663 1455.39792  -0.010
originMH                                -2.29066    2.03149  -1.128
originMI                                -3.10159    2.03678  -1.523
infectionspz:species.molc               14.55447 2301.18562   0.006
infectionuninfected:species.molc        13.29601 1455.39806   0.009
infectionspz:species.molg               29.92816 2058.24342   0.015
infectionuninfected:species.molg        11.97897 1455.39833   0.008
infectionspz:originMH                   -0.40547    1.97062  -0.206
infectionuninfected:originMH             0.61668    1.47262   0.419
infectionspz:originMI                   34.23373 1782.49184   0.019
infectionuninfected:originMI             2.27491    1.54292   1.474
species.molc:originMH                    0.43176    2.15095   0.201
```

```
species.molg:originMH                        -0.01193    1.47545  -0.008
species.molc:originMI                         2.78314    2.28897   1.216
species.molg:originMI                         1.49215    1.39587   1.069
infectionspz:species.molc:originMH           16.52210 1029.12365   0.016
infectionuninfected:species.molc:originMH     0.01577    1.70392   0.009
infectionspz:species.molg:originMH                 NA         NA      NA
infectionuninfected:species.molg:originMH          NA         NA      NA
infectionspz:species.molc:originMI          -18.81921 2058.24429  -0.009
infectionuninfected:species.molc:originMI    -1.67882    1.98388  -0.846
infectionspz:species.molg:originMI          -32.74207 1782.49162  -0.018
infectionuninfected:species.molg:originMI          NA         NA      NA
                                           Pr(>|z|)
(Intercept)                                   0.991
infectionspz                                  0.988
infectionuninfected                           0.993
species.molc                                  0.992
species.molg                                  0.992
originMH                                      0.259
originMI                                      0.128
infectionspz:species.molc                     0.995
infectionuninfected:species.molc              0.993
infectionspz:species.molg                     0.988
infectionuninfected:species.molg              0.993
infectionspz:originMH                         0.837
infectionuninfected:originMH                  0.675
infectionspz:originMI                         0.985
infectionuninfected:originMI                  0.140
species.molc:originMH                         0.841
species.molg:originMH                         0.994
species.molc:originMI                         0.224
species.molg:originMI                         0.285
infectionspz:species.molc:originMH            0.987
infectionuninfected:species.molc:originMH     0.993
infectionspz:species.molg:originMH               NA
infectionuninfected:species.molg:originMH        NA
infectionspz:species.molc:originMI            0.993
infectionuninfected:species.molc:originMI     0.397
infectionspz:species.molg:originMI            0.985
infectionuninfected:species.molg:originMI        NA

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 578.09  on 417  degrees of freedom
Residual deviance: 483.69  on 394  degrees of freedom
AIC: 531.69

Number of Fisher Scoring iterations: 14

> m2=update(m1,~.-infection:origin:species.mol)
> anova(m1,m2,test="Chi")
Analysis of Deviance Table

Model 1: choice2 ~ infection * species.mol * origin
Model 2: choice2 ~ infection + species.mol + origin + infection:species.mol +
    infection:origin + species.mol:origin
  Resid. Df Resid. Dev Df Deviance Pr(>Chi)
1       394     483.69
2       399     491.73 -5  -8.0469   0.1537
> summary(m2)

Call:
glm(formula = choice2 ~ infection + species.mol + origin + infection:species.mol +
    infection:origin + species.mol:origin, family = binomial)

Deviance Residuals:
    Min       1Q   Median       3Q      Max
-1.9258  -1.0285  -0.5927   0.9058   1.9110

Coefficients:
                                Estimate Std. Error z value Pr(>|z|)
(Intercept)                     13.56607  535.41117   0.025    0.980
infectionspz                   -13.19790  535.41330  -0.025    0.980
infectionuninfected            -12.02775  535.41182  -0.022    0.982
species.molc                   -12.50596  535.41145  -0.023    0.981
species.molg                   -12.11677  535.41173  -0.023    0.982
originMH                        -1.85973    1.43990  -1.292    0.197
originMI                        -1.11444    1.47179  -0.757    0.449
```

```
infectionspz:species.molc         11.02764  535.41375   0.021    0.984
infectionuninfected:species.molc  11.70660  535.41217   0.022    0.983
infectionspz:species.molg         11.27033  535.41388   0.021    0.983
infectionuninfected:species.molg  10.76388  535.41238   0.020    0.984
infectionspz:originMH              1.50382    1.13567   1.324    0.185
infectionuninfected:originMH       0.72687    0.73823   0.985    0.325
infectionspz:originMI              1.61505    1.25226   1.290    0.197
infectionuninfected:originMI       1.26021    0.94513   1.333    0.182
species.molc:originMH              0.03303    1.31626   0.025    0.980
species.molg:originMH             -0.70292    1.31078  -0.536    0.592
species.molc:originMI              0.28388    1.27592   0.222    0.824
species.molg:originMI              0.34763    1.20305   0.289    0.773

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 578.09  on 417  degrees of freedom
Residual deviance: 491.73  on 399  degrees of freedom
AIC: 529.73

Number of Fisher Scoring iterations: 12

> m3=update(m2,~.-infection:species.mol)
> anova(m3,m2,test="Chi")
Analysis of Deviance Table

Model 1: choice2 ~ infection + species.mol + origin + infection:origin +
    species.mol:origin
Model 2: choice2 ~ infection + species.mol + origin + infection:species.mol +
    infection:origin + species.mol:origin
  Resid. Df Resid. Dev Df Deviance Pr(>Chi)
1       403     495.66
2       399     491.73  4   3.9301   0.4155
> summary(m3)

Call:
glm(formula = choice2 ~ infection + species.mol + origin + infection:origin +
    species.mol:origin, family = binomial)

Deviance Residuals:
    Min       1Q   Median       3Q      Max
-1.9344  -0.9708  -0.5987   0.9278   2.1687

Coefficients:
                           Estimate Std. Error z value Pr(>|z|)
(Intercept)                 2.26031    0.95659   2.363   0.0181 *
infectionspz               -1.71217    0.92071  -1.860   0.0629 .
infectionuninfected        -0.68194    0.56434  -1.208   0.2269
species.molc               -0.95819    0.85517  -1.120   0.2625
species.molg               -1.33102    0.83125  -1.601   0.1093
originMH                   -2.03824    1.40025  -1.456   0.1455
originMI                   -0.96302    1.42391  -0.676   0.4988
infectionspz:originMH       1.08865    1.04636   1.040   0.2981
infectionuninfected:originMH  0.86533  0.72854   1.188   0.2349
infectionspz:originMI       1.22837    1.20557   1.019   0.3082
infectionuninfected:originMI  1.08847  0.91567   1.189   0.2346
species.molc:originMH       0.04529    1.28774   0.035   0.9719
species.molg:originMH      -0.51918    1.28081  -0.405   0.6852
species.molc:originMI       0.07551    1.22780   0.061   0.9510
species.molg:originMI       0.42551    1.18714   0.358   0.7200
---
Signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 578.09  on 417  degrees of freedom
Residual deviance: 495.66  on 403  degrees of freedom
AIC: 525.66

Number of Fisher Scoring iterations: 4

> m4=update(m3,~.-origin:species.mol)
> anova(m3,m4,test="Chi")
Analysis of Deviance Table

Model 1: choice2 ~ infection + species.mol + origin + infection:origin +
    species.mol:origin
Model 2: choice2 ~ infection + species.mol + origin + infection:origin
```

```
  Resid. Df Resid. Dev Df Deviance Pr(>Chi)
1       403     495.66
2       407     498.01 -4  -2.3455   0.6725
> summary(m4)

Call:
glm(formula = choice2 ~ infection + species.mol + origin + infection:origin,
    family = binomial)

Deviance Residuals:
    Min       1Q   Median       3Q      Max
-2.0593  -0.8993  -0.6727   0.9159   2.0021

Coefficients:
                            Estimate Std. Error z value Pr(>|z|)
(Intercept)                   2.1941     0.7129   3.078 0.002086 **
infectionspz                 -1.6689     0.9124  -1.829 0.067389 .
infectionuninfected          -0.6457     0.5567  -1.160 0.246073
species.molc                 -0.8519     0.5244  -1.625 0.104254
species.molg                 -1.3407     0.5139  -2.609 0.009086 **
originMH                     -2.0885     0.6299  -3.315 0.000915 ***
originMI                     -0.7358     0.8279  -0.889 0.374120
infectionspz:originMH         1.0445     1.0376   1.007 0.314094
infectionuninfected:originMH  0.6956     0.7003   0.993 0.320581
infectionspz:originMI         1.1446     1.2011   0.953 0.340614
infectionuninfected:originMI  1.1801     0.8964   1.317 0.188006
---
Signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 578.09  on 417  degrees of freedom
Residual deviance: 498.01  on 407  degrees of freedom
AIC: 520.01

Number of Fisher Scoring iterations: 4

> m5=update(m4,~.-origin:infection)
> anova(m5,m4,test="Chi")
Analysis of Deviance Table

Model 1: choice2 ~ infection + species.mol + origin
Model 2: choice2 ~ infection + species.mol + origin + infection:origin
  Resid. Df Resid. Dev Df Deviance Pr(>Chi)
1       411     500.41
2       407     498.01  4   2.3997   0.6627
> summary(m5)

Call:
glm(formula = choice2 ~ infection + species.mol + origin, family = binomial)

Deviance Residuals:
    Min       1Q   Median       3Q      Max
-2.0031  -0.9315  -0.6338   0.9649   2.0625

Coefficients:
                   Estimate Std. Error z value Pr(>|z|)
(Intercept)         1.69014    0.58389   2.895  0.00380 **
infectionspz       -0.89306    0.36979  -2.415  0.01573 *
infectionuninfected -0.07748    0.30825  -0.251  0.80154
species.molc       -0.84224    0.51873  -1.624  0.10445
species.molg       -1.33892    0.50965  -2.627  0.00861 **
originMH           -1.45808    0.26040  -5.599 2.15e-08 ***
originMI            0.24905    0.28962   0.860  0.38982
---
Signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 578.09  on 417  degrees of freedom
Residual deviance: 500.41  on 411  degrees of freedom
AIC: 514.41

Number of Fisher Scoring iterations: 4

> m6=update(m5,~.-species.mol)
> anova(m6,m5,test="Chi")
```

```
Analysis of Deviance Table

Model 1: choice2 ~ infection + origin
Model 2: choice2 ~ infection + species.mol + origin
  Resid. Df Resid. Dev Df Deviance Pr(>Chi)
1       413     510.60
2       411     500.41  2    10.19 0.006128 **
---
Signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1
> m7=update(m5,~.-origin)
> anova(m7,m5,test="Chi")
Analysis of Deviance Table

Model 1: choice2 ~ infection + species.mol
Model 2: choice2 ~ infection + species.mol + origin
  Resid. Df Resid. Dev Df Deviance  Pr(>Chi)
1       413     551.22
2       411     500.41  2   50.805 9.284e-12 ***
---
Signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1
> m8=update(m5,~.-infection)
> anova(m8,m5,test="Chi")
Analysis of Deviance Table

Model 1: choice2 ~ species.mol + origin
Model 2: choice2 ~ infection + species.mol + origin
  Resid. Df Resid. Dev Df Deviance Pr(>Chi)
1       413     508.95
2       411     500.41  2   8.5437  0.01396 *
---
Signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

> table(choice2,species.mol)
       species.mol
choice2   a   c   g
  H       6 113 102
  other  21  92  84

> table(choice2,origin)
       origin
choice2  CA  MH  MI
  H      50 137  34
  other  84  49  64
>

> ######### MH vs CA
> x=matrix(c(50,84,137,49),ncol=2)
> chisq.test(x)

        Pearson's Chi-squared test with Yates' continuity correction

data:  x
X-squared = 40.8718, df = 1, p-value = 1.625e-10

> ######### MI vs CA
> x=matrix(c(50,84,34,64),ncol=2)
> chisq.test(x)

        Pearson's Chi-squared test with Yates' continuity correction

data:  x
X-squared = 0.0739, df = 1, p-value = 0.7858

> ######### MI vs MH
> x=matrix(c(137,49,34,64),ncol=2)
> chisq.test(x)

        Pearson's Chi-squared test with Yates' continuity correction

data:  x
X-squared = 39.0592, df = 1, p-value = 4.111e-10

################modele intercept par especes

> m=glm(choice2~1,family=binomial)
> summary(m)
```

```
Call:
glm(formula = choice2 ~ 1, family = binomial)

Deviance Residuals:
   Min      1Q  Median      3Q     Max
-1.734   0.709   0.709   0.709   0.709

Coefficients:
            Estimate Std. Error z value Pr(>|z|)
(Intercept)   1.2528     0.4629   2.706   0.0068 **
---
Signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 28.604  on 26  degrees of freedom
Residual deviance: 28.604  on 26  degrees of freedom
AIC: 30.604

Number of Fisher Scoring iterations: 4

> OR=exp(1.2528)
> OR
[1] 3.50013
> CI1=exp(1.2528+0.4629)
> CI1
[1] 5.560567
> CI2=exp(1.2528-0.4629)
> CI2
[1] 2.203176



> detach(ta)
> attach(t)
> tc=subset(t,species.mol=="c")
> detach(t)
> attach(tc)
> summary(tc)

> m=glm(choice2~1,family=binomial)
> summary(m)

Call:
glm(formula = choice2 ~ 1, family = binomial)

Deviance Residuals:
   Min      1Q  Median      3Q     Max
-1.091  -1.091  -1.091   1.266   1.266

Coefficients:
            Estimate Std. Error z value Pr(>|z|)
(Intercept)  -0.2056     0.1404  -1.464    0.143

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 282.04  on 204  degrees of freedom
Residual deviance: 282.04  on 204  degrees of freedom
AIC: 284.04

Number of Fisher Scoring iterations: 3

> OR=exp(-0.2056)
> OR
[1] 0.8141587
> CI1=exp(-0.2056+0.1404)
> CI1
[1] 0.9368801
> CI2=exp(-0.2056-0.1404)
> CI2
[1] 0.7075125


> detach(tc)
> attach(t)
> tg=subset(t,species.mol=="g")
> detach(t)
```

```
> attach(tg)
> summary(tg)


> m=glm(choice2~1,family=binomial)
> summary(m)

Call:
glm(formula = choice2 ~ 1, family = binomial)

Deviance Residuals:
   Min      1Q  Median      3Q     Max
-1.096  -1.096  -1.096   1.261   1.261

Coefficients:
            Estimate Std. Error z value Pr(>|z|)
(Intercept)  -0.1942     0.1473  -1.318    0.188

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 256.11  on 185  degrees of freedom
Residual deviance: 256.11  on 185  degrees of freedom
AIC: 258.11

Number of Fisher Scoring iterations: 3

> OR=exp(-0.1942)
> OR
[1] 0.8234932
> CI1=exp(-0.1942+0.1473)
> CI1
[1] 0.9541828
> CI2=exp(-0.1942-0.1473)
> CI2
[1] 0.7107035
>




########## Analysis parity #######


> m1=glm(choice2~parity,family=binomial)
> summary(m1)

Call:
glm(formula = choice2 ~ parity, family = binomial)

Deviance Residuals:
   Min      1Q  Median      3Q     Max
-1.254  -1.254   1.103   1.103   1.169

Coefficients:
            Estimate Std. Error z value Pr(>|z|)
(Intercept)  0.01869    0.19336   0.097    0.923
parityP      0.15929    0.24747   0.644    0.520

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 381.69  on 275  degrees of freedom
Residual deviance: 381.27  on 274  degrees of freedom
AIC: 385.27

Number of Fisher Scoring iterations: 3

> m2=glm(choice2~1,family=binomial)
> anova(m1,m2,test="Chi")
Analysis of Deviance Table

Model 1: choice2 ~ parity
Model 2: choice2 ~ 1
  Resid. Df Resid. Dev Df Deviance Pr(>Chi)
1       274     381.27
2       275     381.69 -1  -0.4144   0.5197
> table(choice2,parity)
       parity
```

```
   choice2  N  P
     H     53 77
     other 54 92
```