Environment and Time as Constraints on the Biogeographical Distribution of Gibbons

by

R.I.M. Dunbar

Department of Experimental Psychology, University of Oxford, Oxford, UK

Susan M. Cheyne

Borneo Nature Foundation, Palangka Raya, Indonesia
& Oxford Brookes University, Oxford, UK

Daoying Lan

Guangdong Institute of Applied Biological Resources, Guangzhou 510260, China

Amanda Korstjens

Department of Life and Environmental Sciences, Bournemouth University, Poole, UK

Julia Lehmann

Department of Life Science, University of Roehampton, London, SW15 5JD, UK

G. Cowlishaw

Institute of Zoology, Zoological Society of London, Regents Park, London NW1 4RY, UK
Abstract

We develop a time budget model for the hylobatid family with the aim of assessing the extent to which their contemporary and historical biogeographic distributions might be explained by ecological constraints. The model uses local climate to predict time budgets, and from this the limiting size of social group that animals could manage at a given location. The model predicts maximum group sizes that vary between 3-15 within the taxon’s current distribution, indicating that the combination of their dietary and locomotor styles with the kinds of habitats they inhabit radically constrain group size. Beyond the edges of their current distribution, sustainable group size rapidly tends to zero, although if they had been able to bypass some of these areas, they would have found very suitable habitats in southern India and across the Wallace Line. While travel time would be a major constraint on group size at larger group sizes, as it is in great apes, the main factor limiting the gibbon’s current distribution is the time they need to spend resting that is imposed on them by the environment. The model also indicates that gibbons would not now be able to survive in regions of central and southern China where they are known to have occurred within historical times, perhaps suggesting that historical climate change following the Little Ice Age of the C18th made these regions uninhabitable for them. Finally our results indicate that gibbons have the ecological capacity to live in larger groups than they do, making it unlikely that their adoption of monogamy reflects purely ecological constraints.

Key Words: climate, biogeographic distribution, group size, foraging ecology
Highlights:

- Time budgets limit the biogeographic distribution of gibbons and siamang
- Time that has to be allocated to resting seems to be the main constraint
- The limits on group size are not so restrictive as to make monogamy obligatory
Introduction

Inability to satisfy nutritional and other demands within a defined time period (usually a 24-hr physiological cycle: Peters 1983) limits both where a species can live and, in social species, how big their groups can be (Dunbar, Korstjens & Lehmann 2009). In this respect, time becomes a central problem for all animals, especially for those, like primates, that are not active 24 hours a day. This insight has led to the development of a series of time budget models designed to understand the role of time in limiting the biogeographical distributions of individual species (Dunbar et al. 2009). These models are premised on the fact that behaviour (as the outcome of core physiological processes) is the interface between the environment and the animals’ ability to survive (with the latter indexed as the size of group it can maintain as an ecological entity) (see also Marshall, Carter, Rowcliffe & Cowlishaw 2013). There is little that most animals can do to alter their physiology in response to changes in climate or vegetation, but they can, and do, change their behaviour and group sizes in response to changing environmental conditions.

For species like anthropoid primates that are strictly diurnal due to poor night vision, the time available for satisfying their nutritional requirements is further reduced to the ~12 hours of tropical daylight. In addition, climate and the species’ dietary adaptations may force animals to rest (‘enforced rest’ sensu Korstjens et al. 2010), thereby further reducing the length of their active day. The climatic component of this relationship reflects the fact that almost all tropical animals are obliged to reduce activity and seek shelter during the hottest part of the day when ambient temperatures in open sunlight exceed their thermoneutral zone (Mount 1979; David-Barrett & Dunbar 2016); the dietary aspect is mainly a problem for folivores for whom gut fermentation of foliage is incompatible with any form of activity because...
the bacterial activity on which it depends is extremely sensitive to the heat generated by activity (van Soest 1982).

Intensely social species, like most anthropoid primates, face an additional problem because the functionality of their relationships, as well as the cohesion of their social groups, depends directly on the time they invest in their relationships with each other (Lehmann, Korstjens & Dunbar 2007a; Pollard & Blumstein 2008; Dunbar & Shultz 2010; Dunbar & Lehmann 2013; Sutcliffe, Dunbar, Binder & Arrow 2012; Dunbar 2018a,b). For both humans and nonhuman primates, an individual’s willingness to give coalitionary aid to another depends directly on the amount of time they spend engaged in affiliative interaction (conventionally, social grooming) (Seyfarth & Cheney 1984; Dunbar 1980, 2018a,b; Burton-Chellew & Dunbar 2015).

The need to invest significant quantities of time in social interaction will thus further constrain their time budget flexibility. One implication of this is that animals have to balance the time they invest in essential activities according to the demands set by their environment. At some point, animals will inevitably run out of time, and this will ultimately limit where they can live.

In effect, animals face a three-way optimisation problem in which they have to offset the costs of living in social groups (i.e. time investment in social grooming) and the acquisition of nutrients (specified by the ecological determinants of time required for feeding and travel, including the fixed costs of fuelling fertility and lactation) against the benefits of group size as their principal form of predator deterrence. Predation risk will depend on whether the animal is terrestrial or arboreal and on the hunting styles of the local predators (the ‘landscape of fear’) (Shultz et al. 2004; Shultz & Finlayson 2010; Tolon et al. 2009; Laundré et al. 2014; Coleman & Hill 2014; Riginos 2015; Gallagher et al. 2017), but within these constraints the prey
species can adjust group size to offset local predation risk.

Focusing on time budgets is not an alternative to the more conventional socio-ecology approaches that typically seek to identify correlations between, on the one hand, group size, population density or behaviour and, on the other hand, climate or forage quality/distribution (Wrangham, Gittleman & Chapman 1993; Chapman 2000; Snaith & Chapman 2007; Clutton-Brock & Janson 2012), or those that explore the efficiency of foraging (e.g. Brockleman et al. 2014). Rather, a focus on time budgets provides a way of looking at the mechanisms that underpin the correlations between environmental variables and population level outcomes (group size and dispersion). Time budgets thereby provide insights into the mechanisms involved in these processes, and in particular emphasise a component of that mechanism (namely, time) that is invariably overlooked in most studies.

In addition, a time budget approach allows us to incorporate, directly or indirectly, a range of other factors and causal relationships that bear on animals’ ability to survive in a particular location. This is important because biological phenomena are naturally systems-based, with most causal relationships being multivariate and subject to the influence of feedback loops. Failure to incorporate these more complex relationships may result in the over- or underestimation of the importance of particular causal relationships. Time budget models are allow us to integrate a range of variables and relationships into a single, coherent model.

The formal structure of time budget models is that of a causal chain. The climate at a given location directly or (via its effect on vegetation quality) indirectly determines the three core elements of the time budget (feeding, moving and minimum or ‘enforced’ rest time), and these in turn determine the limiting size of group that animals can maintain at that location. The maximum possible group size predicted by
the time budget ultimately determines whether or not the species can live at that
location (Dunbar et al. 2009). It is important to appreciate that these models are
location-specific: they make explicit predictions about the behaviour of animals at a
given location, subject to the particular climatic and vegetational conditions pertaining
at that location. They are also necessarily genus-specific because they reflect the
particular dietary, physiological and body size adaptations exhibited by a genus. In
this respect, a genus is an ecological species.

Climate variables affect both energy demand (energy animals need to maintain
thermoregulation: Mount 1979) and the rate of nutrient intake through the effect
climate has on the quality and digestibility of forage (and hence the amount of forage
that has to be eaten to extract the required amount of nutrients: van Soest 1982). They
also influence both travel time (through their influence on foraging patch size and
inter-patch distances: Janson & van Schaik 1988; Chapman & Chapman 2000) and
rest time (when animals are forced to seek shelter because ambient temperatures
exceed their thermoneutral zone: Mount 1979; Dâvid-Barrett & Dunbar 2016). In
addition, the feedback loops that are invariably an important component of biological
systems can arise when group size, in particular, influences travel time (because larger
groups have to travel further if feeding patches are depleted easily – typically, less of
a problem for folivores), and this in turn requires more time to be devoted to foraging
to replace the energy consumption of the additional travel.

Time budget models have been published for a number of terrestrial (gelada:
Dunbar 1992a; baboons: Dunbar 1992b; Bettridge, Lehmann & Dunbar 2010; African
great apes: Lehmann, Korstjens & Dunbar 2007b, 2008a) and arboreal (spider
monkeys: Korstjens, Verhoeckx & Dunbar 2006; colobins: Korstjens & Dunbar 2007;
 vervets: Willems & Hill 2009; guenons: Korstjens, Lehmann & Dunbar 2018;
orangutans: Carne, Semple & Lehmann 2012) primates, as well as one ungulate (feral goats: Dunbar & Shi 2013). Time budget models consist of a set of simultaneous equations, one for each of the core time budget variables and have the form of a linear programming (or linear optimization) model (Dunbar 2002). The intersection of these equations defines the limit on group size that a species can maintain at a given location. If any of these relationships involve nonlinear components, solving the simultaneous equation set to find the maximum group size usually requires numerical rather than analytical methods (Dunbar 1992a; Dunbar et al. 2009). However, if all relationships are linear, then analytical methods based on substitution can be used. In the resulting multidimensional state space defined by the climate parameters (the primary drivers) and group size (the final output), the isosurface defined by points where maximum group size drops below the minimum required for demographic viability and/or predator defence defines the taxon’s biogeographic distribution.

These mechanism-based models are at least as successful as conventional climate (or niche) envelope models at predicting the geographical distributions of the taxa concerned (Willems & Hill 2009; Korstjens et al. 2010). In addition, they provide insights into which aspects of the animals’ ecology are responsible for limiting their distribution (Dunbar et al. 2009). This is because time budget models allow us (a) to determine how much ecological and demographic stress a particular population is under (Dunbar 1992a; Lehmann et al. 2007a) and (b) to specify why the species is unable to occupy particular habitats in ways that directly reflect their physiological adaptations (Dunbar et al. 2009). Neither of these is possible with conventional climate envelope models, which are essentially simple correlational relationships. In addition, because time budget models are driven directly by climate, they can be used to explore the consequences of past and future climate change for a taxon’s
biogeographic distribution in a more nuanced way (Lehmann et al. 2010; Kortsjens, 
Lehmann & Dunbar 2010; Bettridge & Dunbar 2012).

Although models have been developed for all three great apes (Lehmann et al. 
2007b, 2008a,b; Carne et al. 2012), the small apes (the gibbon family, Hylobatidae) 
have yet to be considered. The small apes offer a particularly interesting challenge for 
several reasons. First, for an ape, they occupy an unusually exclusive arboreal 
terminal branch niche otherwise occupied only by some of the smaller monkeys. 
Second, they typically live in very small groups (monogamous pairs plus dependent 
offspring) and this raises questions as to why they have opted for this form of social 
system. One classic explanation for monogamy in mammals is that females are forced 
to forage on their own in territories that are too large for a male to be able to 
successfully defend more than one female (Wittenberg & Tilson 1980; van Schaik & 
Dunbar 1990; Komers & Brotherton 1997; Lukas & Clutton-Brock 2013); another 
more recent suggestion is that male parental support allows females to reduce their 
energy demand, especially in strongly seasonal habitats or where locomotor costs 
prohibit fat storage (Heldstab, van Schaik & Isler 2017). Third, they have a relatively 
limited distribution in southeast Asia, and this limited range begs explanation.

Historically, hylobatids probably evolved as a distinct family in southwest 
China in the late Miocene, and gradually extended their range down through the 
Malayan peninsula into the islands of the Sunda shelf as opportunities provided by 
changes in climate and sea level allowed (Jablonski & Chaplin 2009). Turvey, Crees 
& Di Fonzo (2015) found that, as late as the 1800s, gibbons were recorded in 
historical documents as still being widely distributed throughout southern and central 
China as far north as latitude ~35°N (Shanxi, Shaanxi and Shangdong provinces). 
What explains their disappearance from all but the southwest corner of China
(Yunnan and Hainan) remains to be determined, though anthropogenic factors have
been suspected (Fan 2017).

In this paper, we have five main aims. The first is to develop a model of
hylobatid time budgets in terms of climatic variables. Second, we test the model’s
validity by using it to predict the taxon’s presence and absence in various locations
within continental Asia and the islands of the Sunda Shelf and New Guinea. Third, we
use the model to determine which time budget variables are most responsible for
limiting the taxon’s biogeographic distribution. Fourth, we use the model to determine
the possible causes of the historical extinction of gibbon populations in China.
Finally, we use the results to assess how plausible it is that female spacing best
explains hylobatid monogamy.

Methods

A database was compiled from the gibbon literature to provide quantitative
data on key demographic, environmental and ecological variables. We identified 77
studies that provide data on at least some of these (see Dataset S1). Of these, 59
provide data on mean group size, 41 provide data on group density or biomass, and 29
provide data on activity budgets and other behavioural and demographic variables. In
respect of activity budgets, we identify five mutually exclusive states (feeding,
moving or travel, resting, social interaction – mainly social grooming – and singing)
which, in this sample, between them account for ~100% of time.

Inevitably, a number of issues arise when comparing data from different
studies. One is that studies vary in the definitions they use for individual behaviour
categories; a second is that methods of collecting time budget data vary, both in
intensity (e.g. some researchers use group scans, others individual sampling) and in
coverage (the hours of daylight may not be sampled evenly). While these are certainly
issues, past experience developing time budget models suggests that the impact of
these methodological issues is at best modest (see Dunbar & Shi 2013). More
importantly, the main consequence of variations in definition and procedure is that
they increase the error variance in parameter estimates; increased error variance will
simply make it harder to obtain significant results, and will therefore bias the
statistical analysis in favour of the null hypothesis (i.e. no relationship). Ultimately, of
course, the real test is whether our models predict what we see on the ground despite
these methodological flaws. We should be less concerned with standardisation of
definitions or methods (though these should always be encouraged) than with whether
models based on them correctly predict what we know to be the case.

Rainfall, altitude and latitude for each study site are those provided by the
cited publications. Since latitudinal effects should be symmetrical about the equator,
we transform all latitudes into absolute latitude. Individual studies often do not
provide all the indices we need, so we have sourced all temperature variables (mean
annual temperature, TEMP; mean minimum temperature, TEMP\textsubscript{min}; mean maximum
temperature, TEMP\textsubscript{max}; mean daily temperature variation, TEMP\textsubscript{var} [difference
between mean minimum and mean maximum temperature]; and the standard
deviation of mean monthly temperature, T\textsubscript{moSD} [an index of seasonality]) from either
http://www.globalspecies.org/weather_stations/ (which provides climate data for
individual wildlife reserves) or http://en.climate-data.org (which provides climate data
for civic weather stations). Wherever possible, we gave preference to the first. These
climatic variables and the two geographical variables (i.e. latitude and altitude, both of
which are determinants of climate) constitute the set of climatic variables that we will
use as the main independent variables in model-building.
Previous analyses of weather station data for sub-Saharan Africa have demonstrated that only three climatic variables are needed to predict evapotranspiration (the principal predictor of tropical primary productivity: Rosenzweig 1968; Le Houérou & Hoste 1977; Lo Seen Chong, Mougin & Gastellu-Etchegorry 1993). These are mean ambient temperature, total annual rainfall and an index of seasonality (Williamson & Dunbar 1998; see also Hill 1999). As in our previous models for African primates, the standard deviation of mean monthly temperature (TmoSD) is our main index of seasonality. This also allows us to include any effects due to the fact that, in some contexts, certain times of the year when resource availability is poor or climate especially challenging may impose limits on what animals can do (see, for example, Dunbar & Shi 2013).

Where these were given by the original field sources, we also extracted data on a number of ecological and behavioural variables: the percentage of leaf in the diet (%Leaf), percentage of fruit in the diet (%Fruit), the density of gibbon groups (groups/ha), gibbon population biomass (kg/km²), mean day journey length (km), mean territory size (ha), density of fig trees (Figdens, indexed as trees/ha) and number of sympatric primate species. Mean day journey length and time spent moving (transformed into hours spent moving per day) were used to calculate travel speed (m/hr). Together, these comprise the set of ecological and covariate variables we will use in model-building.

For the model-building analysis, we excluded one study on a heavily logged habitat (Johns 1986) because it had an implausibly low estimate for time spent feeding (8%, >3 standard deviations below the mean for gibbons, and almost a full SD below the next lowest value) and an implausibly high value for time spent resting (62%, 2.5 SDs above the overall mean, and a full SD above the next highest value). We also
excluded two studies (West Garo Hills, NE India: Alfred & Sati 1986, 1990; Ujung Kulon, Indonesia: P. Kappeler 1984a,b) that had very high values for time spent feeding (>60%, 3.1 and 3.9 SDs above the mean for gibbons, and >1 SD above the next nearest value) and unusually low values for resting time, as well as one (Tanjong Triang: Ellefson 1974) that had an unusually high value for moving (47%) and low value for rest (7%). Such high values for feeding and moving and low values for resting are indicative of gibbons that were not fully habituated.

Data for all sites are provided in online Dataset S1.

In developing any model, we face a choice between detail and generality. It is always possible to build a model that takes every conceivable environmental variable into account, and hence is 100% accurate in its fit to the data. But in doing so, we inevitably lose generality: in order to make predictions about where the taxon can live, we will need to know much more about the particular environmental parameters of an individual location (e.g. soil type and acidity, inclination and tree composition, as well as all the larger scale variables like rainfall and temperature). Using climate as the basis for a model allows us, at the expense of some loss of accuracy, to be more general, since relevant climate values can be mapped geographically on a continental scale from climate models, as well as being projected backwards and forwards in time.

For the purposes of building a time budget model, we need to determine taxon-specific equations for just two key variables (feeding and moving time). The other two main components (enforced resting time and social time) are derived from general primate equations given, respectively, by Korstjens et al. (2010) and Lehmann et al. (2007a). For the purposes of the present model, we shall take time devoted to singing as a constant (at the mean observed value of 4.96%).
Enforced resting time, as defined by Korstjens et al. (2010), specifies the minimum amount of time that has to be spent resting as a consequence of high ambient temperatures (to avoid thermal overload or excessive heat loss) and for the digestion of leaves (as a function of climatic conditions and the species’ dietary physiology). Enforced resting time differs from observed resting time (time spent inactive, as recorded by field observers) in that observed rest time consists of enforced resting time plus uncommitted time ("free rest") (Korstjens et al. 2010).

Summing the climate-derived time requirements for feeding, moving and enforced rest (plus the constant for singing) tells us how much time is left over in the day that could be devoted to social interaction (grooming). Because grooming time correlates with group size (Dunbar 1992; Lehman et al. 2007; Dunbar & Lehmann 2014), this gives us the maximum size of group that the animals could maintain as a coherent social entity at that location (subject to any feedback loops in the equation set). This does not mean the animals have to devote all this time to social interaction: it simply sets the upper limit.

The social time equation reflects the fact that social grooming is the principal bonding mechanism for primate social groups and increases linearly with group size across primates as a whole (Dunbar 1991; Lehmann et al. 2007; Dunbar & Lehmann 2014). As such, it represents the time investment in social interaction that is necessary to maintain a group’s cohesion through time so as to prevent it disintegrating (Dunbar 2012). It is worth noting that this increase in time devoted to grooming with group size does not mean that animals in large groups groom with more individuals. Typically they do not: rather, as group size increases, animals invest more heavily in the handful of core grooming partners that act as their primary coalition partners within the group (Kudo & Dunbar 2001; Dunbar 2012, 2018a).
As in previous models (for an overview, see Dunbar et al. 2009), we first examined all relevant pairwise plots visually to search for factors that correlate linearly or nonlinearly with the main time budget variables. There were no nonlinear relationships. We then ran backward stepwise linear regression models with the full set of climatic and ecological variables as predictors of each of the two dependent variables (feeding and moving time). The criterion used in selecting the final equation in each case was based on maximising the proportion of variance explained, subject to the requirement that the overall equation and all its individual predictor variables were significant. Because siamang are more than twice as large as other gibbons, siamang feeding time was corrected to a gibbon-equivalent by rescaling it by the ratio of the metabolic body weights of the two taxa (i.e. $5.5^{0.75}/10.75^{0.75}=0.605$). This allows us to use all the data without having to include an additional factor for taxon or body mass. None of the other gibbon species differed significantly in body weight (see Dataset S1).

In extracting these equations from the data, we have ignored the possible influence of phylogeny. As in previous analyses (see Dunbar et al. 2009), we do so principally because almost all of the variables we are concerned with are behavioural rather than anatomical, and are likely to be influenced more heavily by local environmental conditions than by biological inheritance and thus typically have phylogenetic signals that are close to zero (Kamilar & Cooper 2013). In effect, we treat all the hylobatids as belonging to a single ecological species. Inter-population analyses of just this kind have previously been successfully used to explore aspects of the behaviour of gibbons (song function: Cowlishaw 1992, 1996). It is perhaps important to note that we are not seeking to describe the “typical” behaviour of individual gibbon species: rather, our concern is to understand the overall pattern of
behavioural flexibility and the effects of climate on biogeography across populations
of all the hylobatid taxa. For convenience, we adopt the convention of referring to the
entire group as hylobatids, and differentiate between siamang and all other gibbon
species only on the basis of body size.

Collinearity between the climate variables in the field site dataset is, in
general, low and within conventionally acceptable levels, the only exception being
that between rainfall and temperature (mainly due to the effect of the cold, dry
habitats in southwest China disproportionately influencing what is otherwise a much
less clear cut relationship). At best, this relationship explains only 20% of the
variance, and as little as 10% if the Chinese habitats are excluded.

For resting time, we use the generic equation developed for enforced rest
(Rest_{Enf}) by Korstjens et al. (2010) for primates in general. Similarly, for social time,
we use the generic equation developed by Lehmann et al. (2007a) for Old World
primates. Since this equation included variables (sex ratio and female dispersal) that
are constants in the present case, we recalculated a new equation relating social time
directly to group size using the data given by Lehmann et al. (2007a). The equations
are given in Table 1.

We then use this set of equations to determine maximum group size for any
given location. To do this, we sum the feeding, moving and enforced rest times
dictated by the local climatic and habitat conditions, add the 4.96% constant for
singing (the mean time devoted to singing in the sample populations: see Dataset S1),
and ask what time is left over for bonding social groups. The maximum possible
group size can then be determined by interpolating the value for the remaining
uncommitted (or free) time into the generic equation relating social group size to time
devoted to social grooming (given in Table 1). Formally:
$N_{\text{Max}} = \frac{(100 - (\text{Feed} + \text{Move} + \text{Rest})_{\text{Enf}})}{\text{grooming equation}}$

This value is the limiting group size (the maximum ecologically tolerable group size, $N_{\text{max}}$). Since all the equations in the present case are linear, we use analytical methods and solve directly by substitution.

In order to exhibit the biogeographical implications of the model in graphable form, we present the main results as 3-dimensional surface plots of maximum group size plotted against (a) absolute latitude and altitude and (b) mean annual rainfall and mean annual temperature. Because a relatively large number of environmental and behavioural variables are involved in the basic model, we need to convert all other variables into functions of just the two variables used for each graph. We do this by using backwards stepwise regression modelling to derive equations for these variables from the data given in online Dataset S1. The resulting equations are given in Table 2. Of these, latitude, altitude and temperature are well known to be systematically related.

We test the model by evaluating its ability to predict the presence versus absence of gibbons at a range of locations across the Indian subcontinent, the Indo-China peninsula, Indonesia and the Malay Archipelago, New Guinea and mainland China. For these purposes, we selected a number of locations of known altitude and latitude, which had climatic data available in www.en.climatedata.org. The locations and their climatic and geographical data are given in online Dataset S2. Since these analyses predict where hylobatids can and cannot survive, they also allow us to make inferences about the biogeographic distribution of the taxon. In most cases, we selected major cities for these purposes, since our question is whether or not gibbons could live in the general area, not whether they live at a particular location. These data also allow us to assess whether gibbons could now live in those provinces of China.
where they were recorded historically by Turvey et al. (2015), and hence determine whether anthropogenic factors or climate change might have been responsible for their extinction.

**Results**

*The model*

Table 1 lists the best fit equations for feeding and moving that were generated by the data, as well as the generic primate equations for enforced resting time and social time. The causal relationships involved are summarised in Fig. 1. For reference, alternative significant equations for feeding and moving time selected by the backwards stepwise regression are given in the Online Supplementary Material (Table S1). Both feeding and moving time increase as habitats become climatically more stressful (low rainfall, low temperature, high altitude, high latitude). Although group size has no effect on feeding time, it has a weak positive effect on moving time even despite the extremely limited variation in hylobatid group size (Fig. 2). Since moving time is the main limiting factor for great apes, and severely limits their biogeography and group size (Lehmann et al. 2007b, 2008a,b), we ran a separate regression model with just latitude and group size as predictors (Eqn. 2a). It is of significance that, despite an order of magnitude difference in the range of group sizes, the slope coefficient for group size in gibbons is very similar to that for the chimpanzee time budget model (3.08 vs 2.59, respectively).

Across the sample of study populations, percentage of fruit in the diet is determined mainly by ambient temperature (Table 2). In contrast, percentage of leaf in diet increases with the level of environmental stress as reflected in rainfall, altitude and temperature variance (Table 2). Although day journey length was not predicted by...
any of the climatic variables or by fig tree density, territory size is predicted with a very high $r^2$ by a complex of climatic variables (Table 2), suggesting that territory sizes get larger as environmental stress increases.

The next step is to use the time budget equations to predict maximum ecologically possible (i.e. limiting) group size, $N_{max}$, for the sampled populations. To do this, we interpolate the time available for social interaction (i.e. that remaining after removing the time predicted for feeding, moving and enforced rest at the site, plus the constant for singing) into Eqn. (5). For this analysis, siamang predicted feeding time is rescaled back to siamang body mass by reversing the transformation used to convert their feeding time to gibbon-equivalents.

Fig. 3(a) plots maximum tolerable group size, $N_{max}$, against observed mean population group size for all the gibbon and siamang populations in our sample. The horizontal line demarcates a group size of 2.6 individuals, representing the minimum group size for demographic viability (i.e. demographic stability defined by a population growth rate of $r=0$: this requires two surviving offspring over an average gibbon 17-year reproductive lifespan, and would equate to a group with two adults plus $4/17 = 0.24$ births per year to allow for 50% mortality in immatures, hence an average of $\sim 0.6$ surviving offspring aged 0-5 years at any given time). The dashed line marks the line of equivalence ($N_{max}$ is equal to observed mean population size).

Three points may be noted. First, time budget models specify that group sizes should not exceed the predicted maximum size (hence all data points in Fig. 3a should lie above the dashed line). In the present case, 80.5% of the 41 populations have a predicted maximum group size that is larger than their observed mean group size. This is broadly encouraging as far as model fit is concerned, but suggests that something else is needed to account for the exceptions. Second, no population has...
$N_{\text{max}} > 15$ individuals. This suggests that hylobatids are under rather greater ecological
constraint than many other Old and New World monkeys (and African great apes),
where limiting group sizes are typically $>20$. Third, notwithstanding this, many (but
not all) populations have observed mean group sizes that are well below the maximum
possible. This suggests that at least some hylobatid populations could live in
somewhat larger groups, which because of their size would inevitably be multi-
female.

One likely reason why some populations might have maximum predicted
group sizes below their observed group sizes is the length of the active day. Gibbons
are well known to retire early, often several hours before dusk (Raemaekers 1979;
Chivers 1984; Palombit 1997; Fei, Zhang, Yuan, Zhang & Fan 2017), a behaviour
that is quite unusual for most primates. As a result, the sampled activity budgets are
based on an active day that averages 9.3±0.89 hrs (range 8.1-10.6, N=9 sites), despite
the fact that daylength in tropical habitats is ~12 hrs. Progressively extending the
length of the active day in units of 5% suggests that an active day that is 15% (84
min) longer than the observed mean would be enough to lift all but the three lowest
siamang populations above the demographic viability threshold (Fig. 3b). (This would
increase maximum $N_{\text{max}}$ for the other populations only to ~20.) Unfortunately, data on
the length of the active day are not available for any of the populations with $N_{\text{max}}<2.6$.

However, an extra 84 min would only increase the mean value of day length to 10.7
hrs, virtually identical to the longest observed day length in our dataset (10.6 hrs in a
siamang population: Dataset S1).

We ran a sensitivity analysis of the model using the alternative equations for
feeding and moving time from Table S1. To do this, we altered one of the equations at
a time in the original model from Table 1, keeping all the other equations as in the
original model. We also ran a model combining the alternative moving time equation
with the two most extreme alternative feeding time equations. Predicted maximum
group size for the individual populations in the study site database determined using
these alternative equations are highly correlated with those determined using the
original model (Table S2; Fig. S1). The alternative moving time equation increases
the largest maximum group size from ~15 to ~25, but does not substantively change
any of the results. This suggests that the Table 1 model is quite robust. Similarly,
decreasing the slope coefficient for group size in the moving time equation to 2.59
(the value for the chimpanzee model) increases the largest $N_{\text{max}}$ to ~18, while
increasing it to 5.0 decreases $N_{\text{max}}$ to ~9, but does not alter the broad pattern or the fact
that mean population group size is typically comfortably below $N_{\text{max}}$.

To provide some indication as to how environmental parameters influence
hylobatid biogeography, Fig. 4 plots predicted maximum group size as a function of
(a) latitude and altitude and (b) rainfall and temperature. To produce these graphs, we
have used the equations given in Table 2 to reduce all the climatic variables in the
model to the two indices of interest in each case. This can be expected to increase
error variance, with a consequential tendency for reversion to the mean, but the results
provide us with an indication of how maximum group size is likely to vary across
habitats. The results suggest that hylobatids do best (i.e. are able to sustain larger
groups) in cooler, high altitude, high rainfall habitats at low latitudes (i.e. near the
equator). When rainfall is less than ~1500mm per annum, or at latitudes above ~20°
(i.e. outside the Tropics), hylobatids are unable to maintain minimally viable groups
($N$~2.5 individuals) unless they are living at altitudes above ~1000m (i.e. under cooler
conditions) (see also Turvey et al. 2015). This last prediction is confirmed by the fact
that the northern Chinese populations are at significantly higher altitudes than all the
other hylobatid populations (mean altitudes: 1750.0±783.8 SD m vs 401.2±409.5 SD m asl, N=14 and 50 respectively; $F_{1,62}=76.2$, $p<0.0001$).

Testing the model

To explore the model’s ability to predict hylobatid biogeography, we used a sample of locations within the current hylobatid biogeographic range in southeast Asia and a sample of locations on the Indian subcontinent, mainland China and the islands of the Malay archipelago outside the current hylobatid range. In addition, we have included one representative site in each of the Chinese provinces where gibbons were recorded as occurring prior to 1800 AD (see Turvey et al. 2015). Note that, for the latter cases, the climate data are current, not historical, values. In each case, we predict maximum group size using current local climate.

Fig. 5 plots the mean and range of maximum group size predicted by the model for each of these sites grouped by geographical location. Predicted maximum group sizes average 6.6±4.0 SD for 18 sites within the current hylobatid range. In contrast, sites in northeast India and Bangladesh to the west of the Brahmaputra (which forms the northwest boundary of the current gibbon range) never exceed 0.3±0.5 SD (N=7 sites), as is the case for locations further west in northwest India (mean 1.6±2.2 SD, N=7 sites). This suggests that, even in the absence of the physical barrier of the Brahmaputra river system, the taxon’s ability to expand its range westwards has been limited by substantial tracts of land it would have been unable to cross. Had gibbons been able to bypass these two barriers, they would have found very congenial habitats in the southern parts of the subcontinent (e.g. the Nilgiri Hills and adjacent ranges in the Deccan, and in Sri Lanka), where they would have been able to support groups as large as 8.7±5.1 SD (N=7 sites). The island habitats across
the Wallace Line (specifically, the Celebes and Papua New Guinea) provide equally
gibbon-friendly habitats (mean group size 10.3±1.1SD, N=2 sites), but it seems that,
as with most other primates (Brandon-Jones 1998), the gibbons were never able to
traverse this major sea barrier. Increasing the length of the available day by 15% (as
in Fig. 3b) does not change the overall pattern, or make regions such as northern India
or central China any more habitable for gibbons.

More puzzling, however, is the fact that central and east Java (including
nearby islands like Bali) provide very suitable habitats (mean $N_{\text{max}}$ 10.4±1.2SD, N=7
sites) yet have no gibbon populations. Climatically at least, these habitats seem to be
at least as suitable as those in west Java where gibbons do occur (mean $N_{\text{max}}$
10.2±1.9SD, N=7 sites). In contrast, gibbons would now find it impossible to survive
in either the central and southeastern provinces of China where they were recorded as
living prior to 1800 AD (mean 0.0±0.0SD, N=14 sites) or, perhaps less surprisingly,
the provinces of northern China where they did not occur historically (mean
1.3±3.6SD, N=7). This contrasts with locations in southwest China (Yunnan province
and Hainan Island) that currently do support gibbon populations, for which the model
predicts viable group sizes (mean $N_{\text{max}}$ = 4.3±1.7SD, N=3 sites).

What limits hylobatid distribution?

An important feature of time budget models is that they allow us to determine
which aspect of the animals’ biology is the principal constraint on their ability to
occupy habitats, and hence what actually limits their biogeographic distribution. In
turn, this tells us something about the taxon’s risk of extinction under climate change.
To explore this, we plotted the predicted time required for feeding, moving and
enforced rest for the Dataset 2 locations (Fig. 6). Neither feeding nor moving differ
consistently between regions where hylobatids are present vs not present, suggesting that neither of these is a major constraint (except, in the limit, the effect of group size on moving time at much larger group sizes); in contrast, resting is significantly higher in regions where they are absent and seems to account for the fact that they do not have sufficient time for social interaction to maintain groups of any significant size. It thus seems that the main problem lies in the extent to which climatic variables oblige them to rest.

Fig. 7 plots the ratio of observed feeding and resting time divided by the values predicted by the model equations in Table 1 for those populations in the field study dataset (Dataset 1) for whom predicted maximum ecologically tolerable group size is either less than or greater than 5 (the upper limit for observed mean population group size). For these purposes, predicted rest is enforced rest time, while observed rest is total rest time. Feeding time is close to that predicted in both cases, as should be the case. In contrast, observed rest time is significantly less than the minimum predicted (‘enforced’ rest) in populations where Nmax is less than 5, whereas populations where Nmax exceeds 5 typically have surplus rest time capacity.

Discussion

The time budget model for hylobatids that we develop here is broadly similar in form to the models developed for other primate genera. It works well for contemporary populations, predicting presence in most cases where gibbons occur and absence where they do not (Fig. 5). In other cases, it is clear that physical barriers (e.g. habitat conditions in northeast India or the Wallace Line to the southeast) have prevented gibbons expanding into habitats where they could survive well. It is important to remember that time budget models predict the largest groups that a taxon
can maintain at a given site (in effect, the carrying capacity), not necessarily the actual
size of group. They simply set an upper limit defined by local environmental
conditions. Since, for primates in general, fertility is almost always adversely affected
by increasing group size (van Schaik 1982; Dunbar 2018a; Dunbar, MacCarron &
Shultz 2018a), animals will generally try to minimise group size in any given location
to the extent that this is compatible with the constraints imposed by the local
predation risk (Dunbar et al. 2009; Dunbar et al. 2018a).

The fact that hylobatids commonly live in groups that are smaller than those
allowed by the local ecology has two important implications. First, it implies that,
while the model certainly tells us that gibbons are under greater ecological constraint
than most Old World monkeys and African great apes, the level of stress is not
sufficiently high to force them to live in groups as small as those they actually live in.
While it is true that some populations have time budgets with little or no spare
capacity, many do not (Fig. 7). This conclusion is supported by the fact that, unlike
most other monkeys and apes, hylobatids commonly go to their night rest mid-
afternoon (Raemaekers 1979; Srikosamatara 1984; Palombit 1999; Fei et al. 2017),
implying that they are not under significant time pressure. Second, it suggests that
predation risk must be low for hylobatids; group size evidently does not provide
gibbons with a significant anti-predator advantage, so they can afford to minimise
group size in order to maximise fertility (see Dunbar et al. 2009; Bettridge, Lehmann
& Dunbar 2010; Dunbar, MacCarron & Robertson2018b). Indeed, in comparison to
almost all other primates, cases of predation are conspicuous by their absence from
the gibbon literature (Reichard 2003).

This calls into question the claim that gibbons are monogamous because
females are forced by their ecology to forage on their own rather than in groups (van
Schaik & van Hooff 1983; Rutberg 1985; Komers & Brotherton 1997; Brotherton & Komers 2003; Lukas & Clutton-Brock 2013). Contrary to this claim, it seems that most populations could in fact maintain larger groups (range 5-12 animals), especially in the more southerly parts of the Malay archipelago. Groups of this size could easily include 2-4 reproductive females (across all primates, reproductive females form a very consistent 30-35% of the group: Dunbar et al. 2018a,b). Even if females lived alone, it would be possible to have more of them within a gibbon group’s current territory, and males would have access to more than one reproductive female if they pursued a roving male strategy (much as orangutan males do) (see also van Schaik & Dunbar 1990; Bartlett 2009). This implies, as suggested by van Schaik & Dunbar (1990) and Opie, Atkinson, Dunbar & Shultz (2014), that monogamy must have evolved in response to factors other than ecological constraints.

One of the main benefits of the time budget model approach is that it provides insights into the behavioural and physiological constraints that limit a taxon’s ability to successfully occupy different kinds of habitats. As with previous primate time budget models, feeding time is strongly influenced by predictors of habitat quality: in the hylobatid case, the main determinants of feeding time (rainfall, altitude and latitude, with the latter two both being core predictors of temperature) are all variables that influence both tree species composition and the nutritional quality of vegetation (see also Marshall & Leighton 2006). As food quality falls, animals will need to spend longer feeding to meet their nutritional requirements. In contrast, time spent moving remains largely independent of the climatic variables, being affected only by absolute latitude and group size.

Models for African genera have also noted that moving time tends to be independent of environmental variables and is often close to being a constant (Dunbar...
et al. 2009), mainly because, in order to save time for other more pressing activities, animals respond by increasing travel speed rather than increasing travel time when they need to travel further (Dunbar 1992a). There is some suggestion that gibbons also do this: travel speed increases as climatic conditions deteriorate (i.e. when temperatures are high and latitude is low, although the effect is weak (Fig. S2). However, since there is inevitably a limit on the speed at which animals can travel, at some point animals simply have to devote more time to moving. The substantive issue for day journey length (and hence moving time), at least for non-folivores (Snaith & Chapman 2007), is group size: the area that has to be covered to allow every animal in the group to meet its nutritional requirements is inevitably a linear function of the number of animals in the group, forcing the group to travel further each day (Dunbar et al. 2009). The great apes are especially susceptible to the effects of group size on moving time, and it is this that is mainly responsible for chimpanzees’ fission-fusion form of sociality (Lehmann et al. 2007b, 2008a,b). It is significant that, despite the very limited variance in the size of their groups, the group size slope parameter for the gibbons is similar to that for chimpanzees. For groups as small as those found in gibbons, the impact of this effect will, of course, be modest; but it does mean that when groups are larger than ~10 the impact will rapidly becomes prohibitive (see also Lehmann et al. 2007b [Fig. 7]).

However, it seems to be enforced resting time that is the main constraint for hylobatids (Figs. 6 and 7). Since this is mainly affected by ambient temperature, it might explain why social groups get larger in northern populations (Fig. 8). Ultimately, however, the northern extension of hylobatids is constrained by the fact that feeding and moving time are also positive functions of latitude (and hence, in effect, declining temperatures). A comparison of the slope parameters for latitude in
Tables 1 and 2 indicate that the additive effects of the latitude coefficients for feeding and moving time increase at a combined rate that is ~4 times the savings in terms of the slope for enforced rest (0.363 + 0.462 = 0.825 vs 1.33*0.158 = 0.210), so that time budgets become progressively squeezed as animals occupy increasingly high latitude habitats. At the latitude of Mt Wuliang in Yunnan Province, latitude has added a net 15 percentage points to the time budget.

There is a widespread perception that *Nomascus* populations live in larger groups than all other hyllobatids, hence the fact that they have polygamous groups. Up to a point, this is true. However, this is not true for all *Nomascus* populations. Mean population group size is in fact a cubic function of absolute latitude (Fig. 8). Group size does increase steadily up to around 20°N (roughly the latitude of Chaing Mai in northern Thailand), but then it levels off only to drop precipitately after latitude 25°N (roughly the latitude of Mt Wuliang in southwestern China). This crash at very high latitudes is suggestive of populations living at the limits of their range, and indeed mean group size at these highest latitude populations is only just above the minimum for demographic viability.

It is evident from Fig. 3(b) that siamang incur an additional cost due to the fact that their larger body size imposes a higher feeding time demand. A convincing explanation for their larger body size remains elusive, especially given the fact that their range overlaps that of other gibbons and the two taxa are often sympatric. As a result, however, they are often under significant time budget pressure. There is some evidence to suggest that they make time budgeting adjustments that might be sufficient to reduce this pressure: for the limited sample available, it seems that siamang devote less time each day to both singing and social interaction than gibbons do, and they have a longer active day (Table 3). For this limited sample, the net gain
for siamang is equivalent to increasing the length of the active day by \(~17\%\) (the combined effect of a longer active day adjusted for less time spent socialising and singing) while only having to increase the actual length of the active day by \(~10\%\).

That would be sufficient to just lift the lower siamang populations in Fig. 3(b) into the minimum viability zone.

Gibbons share with the great apes a dietary physiology specialised for frugivory; indeed, their feeding time equation is very similar to that of the gorilla, though their size precludes their being able to survive on low quality folivorous fallback foods in the way gorillas can. Although the larger-bodied siamang exhibits some capacity in this direction (Raemaekers 1979; Palombit 1995), their ability to resort to a heavily folivorous diet is likely to be considerably less than the gorilla’s simply because of their smaller body size. Mast years aside, fruits invariably have a more patchy distribution than leaf, and the travel demands imposed by this may be part of the problem that affects gibbon time budgets, and hence limits group sizes.

Gibbon ecology appears to be quite tightly defined: their geographical range is surprisingly limited. They are currently confined to the Indochina peninsula (bridging out into southwest China) and the associated islands of the Malay archipelago (Sumatra, Java and Borneo). They appear to have been prevented from extending southeastwards by the Wallace Line (despite the fact that the islands beyond would be perfectly habitable for them: Fig. 5), while their capacity to encroach into the Indian subcontinent has been limited by the climate west of the Brahmaputra (Fig. 6). Nonetheless, within their core distribution they seem to be under much less ecological constraint than we might have anticipated given the female-dispersion explanation for their adoption of a monogamous mating/social system.

Historically, gibbons are known to have occurred at much higher latitudes in
China than they do now. The model suggests that current climate in these areas makes these habitats completely unsuitable for gibbons. Although the historical extinction of these northern populations has been attributed to anthropogenic factors (Fan 2017), in fact these habitats would not support gibbons now even if there were no humans living there. There are only three plausible explanations for the disappearance of these populations. One is that the populations in question had evolved novel adaptations to these environments, such that the slope parameters on their feeding and/or moving equations were radically different from those for gibbons from further south, but that humans caused their extermination nonetheless. However, to suggest that these northern gibbons were doing something completely different to all other New and Old World anthropoid primates as well as other gibbons implies that they could not have been conventional primates, and that should be, at best, an explanation of last resort.

A second possibility is that humans had released animals in these areas in order to have them as exotics (something that humans seem to be especially prone to do), even though the habitats were not really suitable for them so that the populations went extinct once they were no longer being replenished (a situation not too dissimilar to that of *Macaca sylvanus* on Gibraltar). This is a possibility, but the fact that gibbons seem to be unable to survive there at all would imply the need for constant replenishment, and this seems unlikely over such a wide area and such a long time period. The third option is that the climate has changed in the past 300 years in ways that would have led to the inevitable extinction of these populations (with or without the assistance of humans).

A potential culprit in the latter respect is the Little Ice Age of 1645-1715, which resulted in average world temperatures falling by ~1°C. In China, this resulted in a significant increase in climatic seasonality. The cultivation of Mediterranean-type
citrus crops was abandoned in Jiangxi Province (one of the provinces that historically
harboured gibbons: Turvey et al. 2015) after the 1750s, despite their having been
cultivated there for many centuries (Reiter 2000). Guangdong (another of the
provinces where gibbons have been documented) experienced a particularly cold, dry
spell, accompanied by an unusually high frequency of typhoon strikes, after the Little
Ice Age (Liu, Shen & Louie 2001). Ameca y Juárez, Mace, Cowlishaw, Cornforth &
Pettorelli (2013) have shown, for mammals as a whole and primates in particular, that
high frequencies of cyclones and droughts correlate with elevated extinction risk
(indexed by the number of taxa classified as “threatened”), with the southeast corner
of China being especially prone to this effect. Turvey et al. (2015) noted that there
was increasing fragmentation of these populations from 1700 onwards (and a marked
upturn in fragmentation from around 1900), with a 50-100 year lag to last reported
occurrence.

A comparison of the distribution of climatic variables in southwest China
(where gibbons live now) and central and southeast China (where they lived
historically) suggests that the only climatic variables in which these two regions differ
significantly are TEMP$_{moSD}$ (standard deviation of mean monthly temperature across
the year, an index of seasonality) and annual rainfall (Table 4). (We do not consider
latitude or altitude since these cannot have changed historically.) This may well
reflect shifting patterns in the latitudinal distribution of the monsoon circulation,
allowing an extension of a more seasonal monsoon climate further into mainland
China since the Little Ice Age. A 65% increase in rainfall and a dramatic reduction in
seasonality would be required to allow gibbons to survive in the southeastern
provinces as well as they currently do in the southwest.

The bottom line seems to be that even if anthropogenic factors have been
important in the extinction of these populations (as both Turvey et al. 2015 and Fan 2017 have suggested), these populations’ sensitivity to anthropogenic factors is likely to have been exacerbated by environmental effects on the animals’ increasing ability to cope as climate changed. The results in Fig. 4, for example, imply that, historically, declining populations at high latitudes are likely to have become locked into mountain-top retreats as climate deteriorated, leading to small, isolated pockets that are inevitably more vulnerable to extinction in the face of environmental shocks (Dunbar 1998; Cowlishaw & Dunbar 2000).

One final puzzle is the fact that gibbons are not found east of the Dieng Mountains in central Java. Kappeler (1984a,b) attributed this to the kinds of forests found in the eastern half of the island, which tend to be more deciduous. Fossil gibbons have been recorded at Trinil, central Java (in the Lower/Middle Pleistocene deposits that produced the hominins: Ingicco, Vos & Hoffman 2014) and at the Ngandong and Gunung Dawung hominin sites in east Java (Storm & Vos 2006) (see also Jablonski & Chapin 2009), so this area clearly did once support gibbon populations. The fact that west Java was one of the refugia for gibbons during the dry phases of the late Pleistocene Ice Ages (Brandon-Jones 1998) makes this all the more puzzling. It may be that anthropogenic factors resulted in their demise, although these are unlikely to have kicked in until historical times.

Acknowledgements

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Fig. 1. Flow chart for the gibbon time budget model, summarising the causal relationships between climatic and behavioural variables based on the equations given in Table 1. Solid lines: positive effects; dashed lines: negative effects.

Fig. 2. Time spent moving plotted against mean population group size. Dashed line is least squares regression line. Source: online Dataset 1.

Fig. 3. (a) Maximum ecologically tolerable group size predicted by the time budget model, plotted against observed mean group size for different gibbon (filled symbols) and siamang (unfilled symbols) populations. (b) Predicted maximum group size allowing for a 15% increase in the length of the active day. The solid line demarcates the minimum group size for demographic viability (2 adults plus 0.6 immatures: see text for details); dashed line in (a) is the line of equilibrium ($N_{max} =$ observed mean).

Fig. 4. Maximum group size predicted by the time budget model for different combinations of (a) latitude and altitude and (b) annual rainfall and mean temperature.

Fig. 5. Mean (±95% CI) maximum group size ($N_{max}$) predicted by the time budget model for habitats within different geographic regions where hylobatids are present and not present. Filled circles: regions where gibbons currently live;
unfilled circles: regions where gibbons do not currently live; grey circle: gibbons historically present before 1800 AD. Dashed line: minimum group size for demographic viability. Present: 18 sites within the current biogeographic distribution of gibbons within Indo-China, Sumatra and Borneo. W Java: 3 sites in west Java where gibbons occur; E Java: 8 sites in Java east of the Dieng Mountains where gibbons do not currently occur; E Indonesia: 2 sites on Celebes and western Papua New Guinea, east of the Wallace Line; NW and NE India: 8 and 7 sites, respectively, in northwest and northeast India (including Bangladesh) where gibbons do not occur; S India: 7 sites in southern India plus Sri Lanka; SW China: 3 sites within current gibbon range in Yunnan and Hainan. SE China: one site from each province in China where gibbons were recorded as being present in historical documents before 1800 AD (N=14 site). N China: one site from each province in northern China (including Tibet) where gibbons were not recorded historically (N=7 sites). Locations and climate data from Table S2.

Fig. 6. Mean (±95% CI) (a) feeding time, (b) moving time and (c) enforced rest time predicted by the time budget model for the regional sites shown in Fig. 5. Moving time does not include an adjustment for group size. Regions and symbols as for Fig. 5.

Fig. 7. Mean (±95% CI) ratio of observed to predicted time spent feeding and resting for populations in the field site dataset (online Dataset 1) for which predicted $N_{max} < 5$ (unfilled symbols) or $N_{max} > 5$ (filled symbols). Predicted rest time is enforced rest time.
Fig. 8. Mean observed group size for individual populations plotted against absolute latitude for the population. The best fit equation (dashed line) is Group size = 3.68 + 0.006*Lat + 0.010*Lat^2 – 0.0004*Lat^3 (r^2=0.173, F_{3,47}=3.28, p=0.029).

Source: Dataset S1.
Table 1. Regression equations for the hylobatid model.

<table>
<thead>
<tr>
<th>Variable</th>
<th>Equation</th>
<th>$r^2$</th>
<th>F</th>
<th>df</th>
<th>p</th>
</tr>
</thead>
<tbody>
<tr>
<td>(1) Feed time (%)</td>
<td>$41.634 - 0.003^{<em>} \text{RAIN} - 0.009^{</em>} \text{ALT} + 0.363^{*} \text{LAT}_{\text{abs}}$</td>
<td>0.501</td>
<td>5.03</td>
<td>3.15</td>
<td>0.013</td>
</tr>
<tr>
<td>(2) Moving time (%)</td>
<td>$17.316 + 0.462^{*} \text{LAT}_{\text{abs}}$</td>
<td>0.166</td>
<td>3.59</td>
<td>1.18</td>
<td>0.074</td>
</tr>
<tr>
<td>(2a) Moving time (%)</td>
<td>$5.297 + 0.290^{<em>} \text{LAT}_{\text{abs}} + 3.080^{</em>} \text{N}$</td>
<td>0.220</td>
<td>1.98</td>
<td>2.14</td>
<td>0.175</td>
</tr>
<tr>
<td>(3) Rest_{\text{ENF}} time (%)$^\dagger$</td>
<td>$-23.24 + 1.33^{<em>} \text{TEMP} + 0.259^{</em>} \text{LEAF} + 6.12^{*} \text{T}_{\text{moSD}}$</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>(4) Social(%)$^\dagger$</td>
<td>$2.968 + 0.109^{*} \text{N}$</td>
<td>0.440</td>
<td>29.90</td>
<td>1.38</td>
<td>&lt;0.0001</td>
</tr>
<tr>
<td>(5) Time budget</td>
<td>$100 = \text{FEED} + \text{MOVE} + \text{REST}_{\text{ENF}} + \text{SING} + \text{SOCIAL}$</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>(5a)</td>
<td>$100 = \text{FEED} + (5.297 + 0.290^{<em>} \text{LAT}_{\text{abs}} + 3.080^{</em>} \text{N}) + (\text{REST}<em>{\text{ENF}} + \text{REST}</em>{\text{FREE}}) + \text{SING} + (2.968 + 0.109^{*} \text{N})$</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>(5b) N_{\text{max}}$^\S$</td>
<td>$= ((100 - \text{FEED} - (5.297 + 0.290^{*} \text{LAT}_{\text{abs}} - \text{REST} - 4.96^{\S}) + 2.968)/(0.109 + 3.080)$</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

$\Omega$ Absolute latitude

$\dagger$ Rest time consists of two components: enforced rest (REST_{\text{ENF}}, imposed on the animal by climatic conditions and its dietary strategy) and free rest (REST_{\text{FREE}}) that represents uncommitted time that can be allocated to any other activity when required (see Korstjens et al. 2010).

$^\dagger$ Recalculated from Lehmann et al (2007a)

$^\S$ Maximum ecologically tolerable group size (the maximum group size that will allow the time budget to be balanced)

$^{\S}$ 4.96 = average percentage of day devoted to singing by gibbons
Table 2. Supplementary equations

<table>
<thead>
<tr>
<th>Variable</th>
<th>Equation</th>
<th>$r^2$</th>
<th>F</th>
<th>df</th>
<th>p</th>
</tr>
</thead>
<tbody>
<tr>
<td>TmoSD **</td>
<td>$10.440 + 0.093 \cdot \text{LAT}_{\text{Abs}} - 0.001 \cdot \text{ALT} - 0.821 \cdot \text{TEMP}$</td>
<td>0.597</td>
<td>17.75</td>
<td>4.48</td>
<td>&lt;0.0001</td>
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<tr>
<td></td>
<td>$+ 0.588 \cdot \text{TEMP}_{\text{min}}$</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Mean temperature ($^\circ\text{C}$)</td>
<td>$27.651 - 0.003 \cdot \text{ALT} - 0.158 \cdot \text{LAT}_{\text{Abs}}$</td>
<td>0.672</td>
<td>60.33</td>
<td>2.59</td>
<td>0.0001</td>
</tr>
<tr>
<td>Altitude (m)</td>
<td>$4145.354 - 147.236 \cdot \text{TEMP} + 0.023 \cdot \text{RAIN}$</td>
<td>0.575</td>
<td>37.14</td>
<td>2.55</td>
<td>&lt;0.0001</td>
</tr>
<tr>
<td>Temp$_{\text{var} \dagger}$</td>
<td>10.196</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Absolute Latitude</td>
<td>$48.732 - 0.004 \cdot \text{RAIN} - 1.142 \cdot \text{TEMP}$</td>
<td>0.580</td>
<td>35.23</td>
<td>2.51</td>
<td>&lt;0.0001</td>
</tr>
<tr>
<td>Rain</td>
<td>$3279.523 - 53.609 \cdot \text{LAT}_{\text{Abs}} + 0.025 \cdot \text{ALT}$</td>
<td>0.286</td>
<td>10.60</td>
<td>2.52</td>
<td>&lt;0.0001</td>
</tr>
<tr>
<td>Temp$_{\text{min}}$ ($^\circ\text{C}$)</td>
<td>$-5.776 + 1.033 \cdot \text{TEMP}$</td>
<td>0.922</td>
<td>623.63</td>
<td>1.53</td>
<td>&lt;0.0001</td>
</tr>
<tr>
<td>Territory size (ha)</td>
<td>$2498.88 + 0.077 \cdot \text{ALT} - 477.830 \cdot \text{TEMP}<em>{\text{min}} + 401.718 \cdot \text{TEMP}</em>{\text{max}}$</td>
<td>0.894</td>
<td>19.60</td>
<td>6.14</td>
<td>&lt;0.0001</td>
</tr>
<tr>
<td></td>
<td>$- 477.03 \cdot \text{TEMP}<em>{\text{var}} - 32.620 \cdot \text{TEMP}</em>{\text{moSD}} - 0.032 \cdot \text{RAIN}$</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Leaf in diet (%)</td>
<td>$99.430 - 1.000 \cdot \text{LAT}<em>{\text{abs}} + 0.009 \cdot \text{ALT} + 7.012 \cdot \text{TEMP}</em>{\text{var} \dagger} + 7.879 \cdot \text{TEMP}<em>{\text{min}} - 9.779 \cdot \text{TEMP}</em>{\text{max}}$</td>
<td>0.611</td>
<td>5.04</td>
<td>5.16</td>
<td>0.006</td>
</tr>
<tr>
<td>Fruit in diet (%)</td>
<td>$-51.27 + 10.65 \cdot \text{TEMP} - 7.71 \cdot \text{TEMP}_{\text{min}}$</td>
<td>0.382</td>
<td>6.19</td>
<td>2.20</td>
<td>0.008</td>
</tr>
<tr>
<td>Active day (hr) $\Omega$</td>
<td>$11.273 - 0.001 \cdot \text{RAIN}$</td>
<td>0.905</td>
<td>19.15</td>
<td>1.2</td>
<td>0.048</td>
</tr>
</tbody>
</table>
** Standard deviation of monthly mean temperatures across the year
† Difference between mean monthly maximum and mean monthly minimum temperature
Ω Gibbons only
Table 3. Time budget differences between siamang and gibbons

<table>
<thead>
<tr>
<th>Variable</th>
<th>Siamang</th>
<th>Gibbons*</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Mean±SD</td>
<td>N</td>
</tr>
<tr>
<td>% fruit in diet</td>
<td>40.7±16.1</td>
<td>6</td>
</tr>
<tr>
<td>% leaf in diet</td>
<td>34.6±11.6</td>
<td>4</td>
</tr>
<tr>
<td>Feed (%)</td>
<td>49.6±6.8</td>
<td>5</td>
</tr>
<tr>
<td>Move (%)</td>
<td>18.0±5.8</td>
<td>5</td>
</tr>
<tr>
<td>Rest (%)</td>
<td>34.4±10.1</td>
<td>5</td>
</tr>
<tr>
<td>Sing (%)</td>
<td>2.0</td>
<td>1</td>
</tr>
<tr>
<td>Social time (%)</td>
<td>3.0</td>
<td>1</td>
</tr>
<tr>
<td>Active day (hr)</td>
<td>10.1±0.3</td>
<td>2</td>
</tr>
<tr>
<td>Net difference in available time†</td>
<td>16.8%</td>
<td></td>
</tr>
</tbody>
</table>

* sites used in time budget analyses only

† Equivalent change in gross time budget achieved by the siamang compared to gibbons
(~3% by reducing singing, ~3% by reducing social time and ~11% by increasing length of active day)
Table 4. Comparison of climatic variables between southwest China (where gibbons currently exist) and southeast China (where they were present historically, but no longer are).

<table>
<thead>
<tr>
<th>Provinces of China</th>
<th>Gibbons present</th>
<th>TEMP</th>
<th>TEMP$_{\text{min}}$</th>
<th>TEMP$_{\text{max}}$</th>
<th>TEMP$_{\text{var}}$</th>
<th>TEMP$_{\text{moSD}}$</th>
<th>RAIN</th>
</tr>
</thead>
<tbody>
<tr>
<td>Southwest</td>
<td>current</td>
<td>20.6</td>
<td>16.5</td>
<td>24.6</td>
<td>8.1</td>
<td>3.8</td>
<td>1973</td>
</tr>
<tr>
<td>Southeast</td>
<td>historical</td>
<td>16.6</td>
<td>12.5</td>
<td>20.9</td>
<td>9.7</td>
<td>8.1</td>
<td>1163</td>
</tr>
</tbody>
</table>

$F_{1,15}$  2.50    1.88    2.61    0.21    19.97    9.22

$p$         0.135   0.191   0.127   0.657   0.0005   0.008
Table S1 provides alternative multivariate equations for feeding and moving time generated by the backwards regression model. Most of the slope coefficients are similar across the alternative equations. However, note that the magnitude of the effect of group size on moving time for Equation (2ab) is considerably less. Nonetheless, the values for both Equations (2a) and (2ab) straddle the observed value for the chimpanzee model.

Table S2 gives the correlation values for maximum ecologically tolerable group size predicted by the Original Model based on Table 1 in the main text and the various alternative feeding and moving time equations given in Table S1. All alternative statistical models yield predicted maximum group sizes that correlate significantly with those predicted by the Original Model based on the equations of Table 1.
Table S1. Alternative regression equations for feeding and moving time for the hylobatid model.

<table>
<thead>
<tr>
<th>Variable</th>
<th>Equation</th>
<th>$r^2$</th>
<th>F</th>
<th>df</th>
<th>p</th>
</tr>
</thead>
<tbody>
<tr>
<td>(1) Feed time (%)</td>
<td>$41.634 + 0.363 \times \text{LAT}_{abs} - 0.009 \times \text{ALT} - 0.003 \times \text{RAIN}$</td>
<td>0.501</td>
<td>5.03</td>
<td>3,15</td>
<td>0.013</td>
</tr>
<tr>
<td>(1a)</td>
<td>$53.955 + 0.243 \times \text{LAT} - 0.010 \times \text{ALT} - 0.472 \times \text{TEMP} + 0.243 \times \text{TEMPVAR} - 0.004 \times \text{RAIN}$</td>
<td>0.340</td>
<td>2.86</td>
<td>5,13</td>
<td>0.059</td>
</tr>
<tr>
<td>(1b)</td>
<td>$55.879 + 0.279 \times \text{LAT}_{tabs} - 0.010 \times \text{ALT} - 0.485 \times \text{TEMP} - 0.004 \times \text{RAIN}$</td>
<td>0.520</td>
<td>3.80</td>
<td>4,14</td>
<td>0.027</td>
</tr>
<tr>
<td>(1c)</td>
<td>$33.230 + 0.469 \times \text{LAT}_{tabs} - 0.009 \times \text{ALT}$</td>
<td>0.461</td>
<td>6.85</td>
<td>2,16</td>
<td>0.007</td>
</tr>
<tr>
<td>(2) Moving time (%)</td>
<td>$17.316 + 0.462 \times \text{LAT}_{abs}$</td>
<td>0.273</td>
<td>5.25</td>
<td>1,14</td>
<td>0.038</td>
</tr>
<tr>
<td>(2a)</td>
<td>$5.297 + 0.290 \times \text{LAT}_{abs} + 3.080 \times \text{N}$</td>
<td>0.220</td>
<td>1.98</td>
<td>2,14</td>
<td>0.175</td>
</tr>
<tr>
<td>(2aa)</td>
<td>$29.196 + 0.306 \times \text{LAT} - 0.365 \times \text{TEMP}_{max}$</td>
<td>0.294</td>
<td>2.71</td>
<td>2,13</td>
<td>0.104</td>
</tr>
<tr>
<td>(2ab)</td>
<td>$19.409 + 0.303 \times \text{LAT}<em>{abs} - 0.275 \times \text{TEMP}</em>{max} + 1.718 \times \text{N}$</td>
<td>0.312</td>
<td>1.81</td>
<td>3,12</td>
<td>0.199</td>
</tr>
</tbody>
</table>
Table S2. Correlations between maximum group size predicted by the original model of Table 1 and a model using the different equations for feeding and moving time given in Table S1, for the actual hylobatid study sites (as listed in Dataset-1). In each case, the remaining equations are as in the original model of Table 1.

<table>
<thead>
<tr>
<th></th>
<th>Feed Eqn 1a</th>
<th>Feed Eqn 1b</th>
<th>Feed Eqn 1c</th>
<th>Move Eqn 2ab</th>
<th>Feed Eqn 1a with Move Eqn 2ab</th>
<th>Feed Eqn 1c with Move Eqn 2ab</th>
</tr>
</thead>
<tbody>
<tr>
<td>Original model</td>
<td>Correlation</td>
<td>0.994</td>
<td>0.995</td>
<td>0.988</td>
<td>0.996</td>
<td>0.987</td>
</tr>
<tr>
<td></td>
<td>p (2-tailed)</td>
<td>&lt;0.0001</td>
<td>&lt;0.0001</td>
<td>&lt;0.0001</td>
<td>&lt;0.0001</td>
<td>&lt;0.0001</td>
</tr>
<tr>
<td></td>
<td>N</td>
<td>51</td>
<td>51</td>
<td>51</td>
<td>51</td>
<td>51</td>
</tr>
<tr>
<td>Feed Eqn 1a</td>
<td>Correlation</td>
<td>0.999</td>
<td>0.974</td>
<td>0.996</td>
<td>0.997</td>
<td>0.978</td>
</tr>
<tr>
<td></td>
<td>p (2-tailed)</td>
<td>&lt;0.0001</td>
<td>&lt;0.0001</td>
<td>&lt;0.0001</td>
<td>&lt;0.0001</td>
<td>&lt;0.0001</td>
</tr>
<tr>
<td></td>
<td>N</td>
<td>51</td>
<td>51</td>
<td>51</td>
<td>51</td>
<td>51</td>
</tr>
<tr>
<td>Feed Eqn 1b</td>
<td>Correlation</td>
<td>0.975</td>
<td>0.997</td>
<td>0.997</td>
<td>0.997</td>
<td>0.980</td>
</tr>
<tr>
<td></td>
<td>p (2-tailed)</td>
<td>&lt;0.0001</td>
<td>&lt;0.0001</td>
<td>&lt;0.0001</td>
<td>&lt;0.0001</td>
<td>&lt;0.0001</td>
</tr>
<tr>
<td></td>
<td>N</td>
<td>51</td>
<td>51</td>
<td>51</td>
<td>51</td>
<td>51</td>
</tr>
<tr>
<td>Feed Eqn 1c</td>
<td>Correlation</td>
<td>0.984</td>
<td>0.966</td>
<td>0.997</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>p (2-tailed)</td>
<td>&lt;0.0001</td>
<td>&lt;0.0001</td>
<td>&lt;0.0001</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>N</td>
<td>51</td>
<td>51</td>
<td>57</td>
<td>51</td>
<td>51</td>
</tr>
<tr>
<td></td>
<td>Correlation</td>
<td></td>
<td></td>
<td></td>
<td></td>
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</tr>
<tr>
<td>----------------</td>
<td>-------------</td>
<td>----------</td>
<td>----------</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Move Eqn 2ab</td>
<td>0.995</td>
<td>0.989</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>p (2-tailed)</td>
<td>&lt;0.0001</td>
<td>&lt;0.0001</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>N</td>
<td>51</td>
<td>51</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Feed Eqn 1a with Move Eqn 2ab</td>
<td>0.977</td>
<td></td>
<td></td>
<td></td>
<td></td>
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</tr>
<tr>
<td>Correlation</td>
<td></td>
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<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>p (2-tailed)</td>
<td>&lt;0.0001</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>N</td>
<td>51</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Fig. S1 (below) plots predicted maximum group size for individual study sites in the main dataset (online Dataset-1) against the equivalent value predicted by substituting the various alternative feeding and moving time equations. Alternative feeding time equations do not have a significant effect on predicted maximum group size. Note that the alternative moving time equation predicts slightly larger group sizes than those predicted by the Original Model of Table 1 because the effect of group size on moving time in Equation 2ab is lower (1.72 vs 3.08), yielding slightly higher group sizes. Nonetheless, the difference in the size of the largest groups predicted is modest (25 vs 15), the outcomes do not differ significantly (Table S2) and using these alternative equations does not change any of the main conclusions.
Figure S1

Maximum group size predicted by the original model using the Table 1 equations plotted against maximum group size using the alternative Feed and Move equations given in Table S1. Except for the last pair of graphs, each alternative equation is substituted individually, and all other equations are held constant as in the Original Model of Table 1. Filled symbols: gibbon populations; unfilled symbols: siamang populations. Dashed line: line of equivalence (the predicted maximum group size is the same for both models).
Fig. S1 (contd)
**Speed of travel**

We calculate average speed of travel across the day as day journey length (in metres) divided by the number of hours in the day (12) multiplied by the proportion of the day devoted to moving. Speed of travel is weakly predicted by latitude, altitude and temperature:

\[
\text{SPEED (m/hr) = -621.2 – 25.7* LAT}_{\text{Abs}} + 0.4* \text{ALT} + 46.5* \text{TEMP}
\]

\[
(r^2=0.674, \ F_{3,7}=1.94, \ p=0.211).
\]

Fig. S2(a) plots speed against absolute latitude and mean temperature.

Fig. 2(a) has a very slight suggestion that the relationship might be quadratic (U-shaped) in temperature, with speed increasing when ambient temperatures are both low and high, with a minimum at temperatures around 20°C. Alternatively, the relationship may be better described by a power curve in which speed increases dramatically above ~22°C at latitudes <5° either side of the equator.

**Figure S2(a)**

Mean speed for individual gibbon (filled symbols) and siamang (unfilled symbols) plotted against mean temperature and absolute latitude of the study site, assuming a 12-hour tropical day.
Many gibbon populations retire to nest several hours before sunset, with length of active daylength best predicted by mean rainfall (Table 2). To check whether a shorter active day makes any difference to these results, we recalculated speed as a function of actual length of active day using the equation given in Table 2 to predict length of active day from rainfall for each site. Aside from a slight uniform increase in speed, the transformation has no effect (Fig. S2b).

**Figure S2(b)**
Mean speed for individual gibbon (filled symbols) and siamang (unfilled symbols) plotted against mean temperature and absolute latitude of the study site, with speed calculated for active day length (estimated using the equation in rainfall from Table 2).
Figure 2

![Figure 2](image)

**Time spent moving (%)**

**Group size**