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1	Hominin home ranges and habitat variability: exploring modern African analogues using remote
2	sensing
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14	Abstract
15	The palaeoanthropological literature contains numerous examples of putative home range sizes
16	associated with various hominin species. However, the resolution of the palaeoenvironmental
17	record seldom allows the quantitative analysis of the effects of different range sizes on access to
18	different habitat types and resources. Here we develop a novel approach of using remote sensing
19	data of modern African vegetation as an analogue for past hominin habitats, and examine the
20	effects of different range sizes on the access to habitat types. We show that when the location of the
21	ranges are chosen randomly then the number of habitat types within a range is surprisingly scale

22 invariant – that is increasing range size makes only a very modest difference to the number of

23 habitat types within an estimated hominin home range. However, when transects are placed

24 perpendicular to a water body (such as a lake or river bank) it is apparent that the greatest number

25 of habitats are seen near water bodies, and decline with distance. This suggests additional

advantages to living by freshwater other than the obvious one associated with access to drinking

27 water, and may indicate that the finding of hominins in fluvial and lacustrine deposits is not simply a

28 taphonomic issue.

30 Introduction

31 In the nineteenth and early twentieth century relatively little emphasis was given to the 32 environmental context in studies of human evolution – this started to change in the 1930's around the time of the 'evolutionary synthesis' (Bowler, 1986). While there is now some consensus in the 33 34 literature that many early hominins in Africa lived in mosaic habitats (see Reynolds et al. (2015) for a 35 review and history of this terminology), little work has been undertaken on how variable habitats 36 might have been within hominin home ranges. While site-level analyses can produce highly detailed 37 results (e.g. Kroll and Isaac, 1984; Magill et al. 2016), and analyses integrating climate and 38 palaeoproxies have been undertaken at a continental scale (e.g. Blome et al., 2012), very little 39 detailed landscape reconstruction has been attempted at the level of individual hominins or their 40 social groups. Here we take a novel approach to hominin spatial ecology by using remote sensing to 41 quantify patterns in the vegetation of modern Africa at hominin-relevant scales, to examine the 42 distribution and habitat variability that may have been encountered in the past. Such an approach 43 has the obvious disadvantage of characterising the modern vegetation, rather than the vegetation at 44 the time of interest for any given past hominin species. However, it does allow variation to be 45 quantified at a far greater spatial and narrower temporal scale than is possible based on 46 palaeoenvironmental proxies such as pollen, pedogenic carbonate analysis or phytoliths (although 47 these can be incorporated, see below), or from traditional field-based approaches. We explore 48 these advantages here, using data from seven separate regions of sub-Saharan Africa to quantify 49 habitat variability at a variety of hominin-relevant scales.

50 We are particularly considering the range sizes and habitat variability associated with various species 51 of Australopithecus and early Homo, although the methods are equally applicable to earlier and later 52 hominin taxa. While it is obvious that hominins lived on and within the landscape, we have few tools 53 at our disposal to examine exactly how different habitat types may have influenced their 54 movements. Suggested key characteristics are the presence of water (Ashley et al. 2009; Finlayson, 55 2014; Quinn et al. 2013), trees for shade (Habermann et al., 2016), and river cobbles or outcrops for tool making (Harmand, 2009). It is rare, however, to have stone tools or cutmarked bones directly 56 57 associated with palaeoenvironmental proxies that can be used to reconstruct that exact location (although FLK Zinj may be a notable exception, Magill et al. 2016). Rather than seeking to 58 59 reconstruct a particular place at a specific point in time (which can usually only be achieved on a scale of tens to hundreds of metres, rather than the kilometres that hominins are likely to have 60 ranged), we are examining vegetation at a larger scale to look for physical patterns – such as the 61 62 number of different habitat types a hominin may have encountered in a daily round, or even within

their lifetime. If the presence of a number of different habitat types such as trees, bushes, water,
swamp or bare rock was important to hominins, then we can examine how likely it is that such
variability is would have been encountered on a regular basis or at specific locations (such as
riversides) using vegetation classifications derived from modern remote sensing.

67 This new approach allows us to consider, for measures identified in the fossil record (e.g. % canopy 68 cover, Cerling et al. 2011; Quinn et al., 2013) and vegetation patchiness (i.e. 'mosaic' habitats), how 69 the land cover in modern Africa varies on a number of hominin-relevant scales. Questions such as 70 'how many vegetation types would typically be found within the putative range size of a given 71 hominin species and how does this number vary as range size increases (or decreases)?'. In this 72 paper we set out the basic ideas of this approach – which is intended to be complimentary to, rather 73 than replacing, existing ways of addressing these questions. We use our data to address two specific 74 points:

We look at randomly placed home ranges and calculate land cover within them. This
allows us to quantify the effect of increasing range size on access to different vegetation types.

2) We focus on water, and examine how land cover and patchiness change as one movesaway from water sources.

Note that we are not attempting to reconstruct past environments, rather quantifying the landscape as it is today (with adjustment for anthropogenic change, see methods) and using this as a surrogate for the unquantifiable spatial variation of the past. We also provide, as an illustrative example of the ways in which this approach could be developed, a brief case study which compares data from our analysis with data gained from pedogenic carbonates from East African hominin localities.

84 Home ranges

85 A home range may be defined as a circumscribed area in which an individual spends much of its life, and contains the requisite resources (food, water, shelter and conspecifics to mate with) (Barnard, 86 87 1999). This is somewhat different to a territory, which is the section of a home range that is actively 88 defended (Manning and Dawkins, 2012). A territory may cover the entire home range or be 89 restricted to around a particular resource, such as nesting site. Barnard (1999) points out that home 90 range size is not always easy to quantify for extant animals, and it is even harder to infer for extinct 91 taxa. However, there are some general rules than can be applied; for example, home range size (or 92 feeding territory size) tends to be scaled with body mass (Clutton-Brock and Harvey, 1984). This is 93 unsurprising as not only do larger animals require more physical space but they also require more 94 food than smaller animals (McNab, 2012).

Following from these patterns established for extant species, for hominins an increase in range size
has been inferred from increasing carnivory (Foley, 2001), a direct result of increasing body size and
dietary quality (Leonard and Robertson, 2000, and see below). However, it has proved difficult to
gain accurate estimates of home range size, even using modern isotopic techniques (e.g. Copeland
et al. 2011), so we have used a variety of measures based on archaeological information and

- 100 estimates based on data from extant human groups or other animals.
- 101

102 Methods

103 Calculating landcover

To quantify land cover heterogeneity in a variety of modern African landscapes, we analysed seven Landsat ETM+ satellite image pairs ranging in latitude from Ethiopia to South Africa, and in habitat types from forest to semi-desert (Fig. 1). These were chosen from a larger study of sub-Saharan Africa land cover, in whichsites were selected randomly and from thesewe chose seven sites that we considered representative of the main habitats and geographical locations most often discussed in studies of early human evolution in Africa.

110 Due to the highly seasonal nature of many African landscapes as a result of climate and rainfall patterns, both wet and dry season Landsat ETM+ satellite imagery was used in combination to 111 112 generate a single land cover classification for each study area. This enabled land cover classes 113 present, or only able to be discriminated, at certain times of the year, such as seasonal water to be 114 identified. A number of image pre-processing steps were performed on the images using ERDAS 115 IMAGINE 2010 to ensure data quality was maintained. These included: error detection and 116 recording; cloud and cloud shadow masking; image geometric accuracy checking; atmospheric correction; and finally compositing the wet and dry season images into a single dual-date composite 117 118 image (Morton et al., 2011). For both the wet and dry season Landsat ETM+ images spectral bands 1 119 (0.45-0.52 μm wavelength), 2 (0.52-0.60 μm), 3 (0.63-0.69 μm), 4 (0.77-0.90 μm), 5 (1.55-1.75 μm) and 7 (2.09-2.35 μ m) were used to enable characterisation of the varying wavelength-dependent 120 121 spectral response of land surface features. Band 6 (thermal, 10.40-12.5 μm) was used only during 122 the cloud masking stage, and was not included in the final composited image. The composite images 123 were projected in the Universal Transverse Mercator (UTM) WGS84 coordinate system.

An unsupervised pixel-based classification technique with post-classification refinement was used to generate land cover maps of the study areas. Unsupervised classification algorithms aggregate all pixels within an image into groupings based on the spectral characteristics of those pixels, with the 127 clustering process controlled by predetermined parameters for numbers of iterations and classes 128 generated (Loveland and Belward, 1997). Unsupervised classification techniques are well established 129 for land cover mapping applications, and have been used in the production of regional and global 130 land cover maps (Loveland et al., 2000; McGwire et al, 1992; Fleischmann and Walsh, 1991). An 131 unsupervised classification generating 75 spectral classes was produced for the composite image 132 using the Iterative Self-Organising Data Analysis Technique (ISODATA) (Bezdek, 1973). This large number of classes was used to minimise the problem of split land cover class spectral clusters 133 134 (Horner et al., 1997; Wayman et al., 2001).

135 High-resolution satellite imagery of the study areas available via public portals such as Google Earth 136 was used a reference to enable the unsupervised spectral classes to be assigned specific land cover 137 class labels (Loveland et al., 2000; Juang, et al., 2004; Cihlar, 2000) corresponding to the project 138 classification nomenclature in Table 1. Additionally, field surveys conducted in the Kruger National 139 Park, South Africa (shown as area F in Fig. 1), in July 2014 involved further identification of ground 140 truthing locations of known land cover types for validation of the classification generated at this site. 141 This field data was combined with the high resolution imagery-derived validation data and showed 142 good congruence between methodologies (Marston et al. in prep.), however given the logistical 143 challenges of collecting ground truthing data over such broad geographical areas, high resolution 144 reference imagery provided the sole source of validation data for the other sites. The classification 145 nomenclature used was designed to be applied broadly across sub-Saharan Africa and was based on 146 a modified version of the Global Land Cover 2000 Land Cover Map of Africa classification system 147 (Mayaux et al., 2000). Our classification also pays special attention to the forest – grassland gradient, and follows the approach of Torellos-Raventos et al., (2013) which stratified this gradient into five 148 149 forest to grassland categories at 25% intervals (100-75%, 75-50%, 50-25%, 25-5% and 5-0%). We 150 have amalgamated the latter two categories to form a 25-0% canopy grouping. The generated 151 classification maintained the 30 m spatial resolution of the input Landsat ETM+ imagery.

Although the unsupervised spectral classes generally corresponded well to specific land cover classes, occasionally they contained groups of pixels that when inspected were found to relate to more than one land cover class. For these areas of known misclassification, post-classification refinement in the form of manual knowledge-based enhancement procedures was performed to split these classes into single category sub-classes (Loveland et al., 2000), and also to re-label land cover patches to resolve the spectral confusion between disparate land cover classes.

158 Multiple unsupervised spectral classes would frequently correspond to the same land cover class in 159 the nomenclature due to the inherent spectral variability of that class. For example, the agriculture 160 class comprised multiple crop types with different planting, growth cycle, harvesting and watering 161 characteristics which are spectrally distinct when present in the satellite imagery. This enabled the 162 spectral variability of each land cover class to be captured prior to the unsupervised spectral classes 163 being aggregated using well defined merging steps (Juang et al., 2004) until a single merged class for 164 each desired land cover class was achieved.

165 Accuracy assessment of the classification using ground truth locations was performed. Ground 166 truthing data (i.e. independent verification) of known land cover types was derived from high-167 resolution imagery of the survey area (e.g. Google Earth), and also from field survey data for site F. 168 The use of higher resolution imagery as a source of validation data for testing the accuracy of 169 classifications derived from coarser resolution satellite products such as Landsat ETM+ imagery is an 170 established technique (Duro et al., 2012; Xie et al., 2008; Cihlar et al., 2003). Due to differences 171 between the Landsat ETM+ and high resolution reference data acquisition dates, all points exhibiting suspected temporal change or human or natural disturbance between the two acquisition dates 172 173 were disregarded. Classification accuracy assessments are shown in Table 2, and confusion matrices 174 for all study areas are provided as Supplementary Information.

Topographical variability across the buffer extents was also examined using Shuttle Radar
Topography Mission (SRTM) digital elevation model (DEM) data. Slope data was derived from the
SRTM DEM data, with mean, minimum, maximum and standard deviation variables for both
elevation and slope extracted for each buffer using the Geospatial Modelling Environment software.
Patch richness for all buffers at each radius size was compared to mean elevation and mean slope
data across the dataset (data not shown), and no convincing relationships were found.

181 Range-size estimates

182 We have used a number of postulated hominin home range sizes from the literature, deliberately 183 sampling a wide size range, supplemented with estimates based on archaeological and 184 anthropological data. Milton and May (1976) proposed a method of estimating home range sizes for 185 individual primates based on body masses and known group-home range sizes. For group-living 186 primates the range size for an individual was estimated by dividing the group home range size by the 187 numbers of individuals within the group. While this necessarily creates an under-estimate of the 188 area any individual may actually roam, it has been widely used and cited in hominin studies (e.g. 189 Leonard and Robertson, 2000; Antón et al. 2002; Antón and Swisher, 2004). Leonard and Robertson 190 (2000) also included estimates of diet quality (whether ape-like or human-forager-like) to calculate 191 their estimated hominin home range sizes. Their equations were subsequently used in Antón et al.

192 (2002), but with slightly different results. Here we have used the figures from Antón et al. (2002), 193 taking the smallest estimated hominin range size (38 ha for Australopithecus africanus on an ape-like 194 diet) and largest pre-sapiens range size (452 ha for Homo erectus on a low quality human forager-195 like diet) (see Table 3). We have also utilised published data on modern Hadza maximum foraging 196 distances (Raichlen et al. 2014). For this home range estimate we used the median distance (~1,100 197 m) based on the maximum distance travelled during 715 foraging bouts (Raichlen et al., 2014). The 198 largest range size is based on the 13 km distance estimated from the original sources of the raw 199 material found at the Oldowan site of Kanjera, Kenya (Braun et al. 2008). While this estimate is 200 necessarily approximate, it provides a useful larger range size and is one of the few approaches 201 available for estimating range size directly from the archaeological record. We also calculated an 202 intermediate home range size estimate of 2.5 km, covering an area of some 19.6km², as an 203 additional model.

204

205 Data cleaning and sample sizes

206 The calculated home range sizes from Table 3 were overlaid onto the classified images as circular 207 areas (buffers). Circular buffers are the simplest geometric shape with which to undertake these 208 types of analysis. While clearly not capturing the complexity of an animal's daily or yearly movement 209 patterns, they have a long history of use in archaeology as a heuristic device (e.g. Vita-Finzi and 210 Higgs, 1970; Flannery, 1976; Bird et al. 2008, Grove, 2009). Our buffers were fixed on a central point, 211 with increasing radii corresponding to the estimated home range sizes (Fig. 2). There were 300 212 randomly located central points per image. Once the buffers had been applied around these central 213 points, the data were quality checked to remove all buffers containing any cloud, >80% saltwater or 214 freshwater or >10% anthropogenic land cover classes (arable agriculture, built-up environments, and 215 coniferous plantations). Any point where buffers extended beyond the classified area at any buffer 216 size were also disregarded, and only the central points that remained across all five radii sizes 217 retained for further analysis, ensuring that the same buffers were being examined at each scale. This 218 left a variable number of buffers in the analysis for each study area (area A, n = 19; area B, n = 82; 219 area C, n = 48; area D, n = 31; area E, n = 78; area F, n = 174; area G, n = 164), totalling 596 buffers 220 for each estimated home range size.

221 Analysis

We calculated patch richness (PR) based on the land cover classifications shown in Table 1. PR is a simple measure of how many land cover types (e.g. open woodland, closed woodland) there are

within each buffer. As we are looking to perform analogous studies of hominin landscapes, for the
PR results presented here we removed all land cover types that are clearly anthropogenic, leaving
only 'natural' vegetation types present (i.e. if a buffer had a PR of 5, but one of the land cover types
was 'built up' we removed it to give a 'natural' PR of 4).

We also calculated the percentage of canopy cover within each randomly placed buffer using four categories. Closed woodland = >75% canopy cover, open woodland = 75% - 50% canopy cover, discontinuous grassland = 50% - 25% canopy cover and continuous grassland = 25% - 0% canopy cover. To make these buffer data directly comparable to the vegetation palaeoproxy data from pedogenic carbonates, where other land covers were present, such as bare or swamp they were disregarded and the four % canopy cover categories were scaled to cover 100% of the buffer.

234

235 Transects

Complementary to the randomly located buffers which examine the general PR and land cover 236 237 variability, targeted higher resolution analysis of the localised areas around rivers was performed. 238 This involved the selection of twenty-one transects in two areas (area B, Kenya, n = 9; area F, South 239 Africa, n=12). Each transect started in and then moved away from a water body or river channel with 240 data extracted every 10 m for 5 km along the transects, with this dense sampling providing highly 241 detailed information on localised landscape variability . Two types of data were extracted for each 242 point along the transect - PR data with the central point of the buffers located on the transect for 4 243 different buffer sizes (347 m, 1100 m, 1199 m and 2500 m), and the land cover class of each 244 individual transect point. The transect locations were selected by eye to cover areas without human 245 activity (such as tarmac roads, agriculture, etc.). The largest 13 km radii buffer size had the effect of 246 smoothing the PR values to a degree where variability in PR values was lost along the transect extent 247 when sampled at 10 m intervals. Therefore the 13 km radii size was disregarded from this element of 248 the analysis which then focussed on the smaller radii. This did not affect the randomly located 249 sample points due to the greater spacing between them. Note that the transect analyses reported 250 here are intended to be illustrative rather than representative of the full range of possible results.

251

252 Results

- 253 Accuracy assessment was performed on the land cover classifications for the seven study areas, with
- high accuracies observed from a minimum of 81.6% (area E) to a maximum of 91.8% (area B)
- recorded (Table 2). Individual site class error matrices are presented in supplementary information.

256 *Randomly placed buffers*

257 Patch Richness

The results for PR are shown in Table 4. Perhaps unsurprisingly it shows that as the buffer sizes get
larger, the number of different habitats within them increases, yet the difference in medians
between the smallest and largest buffers are quite modest. The least variation (increase in median
PR of 1) is seen in area G (South Africa), and the greatest (increase in median PR of 3) in area D
(Rwanda/Burundi) and area F (South Africa/Mozambique).

Table 4 also demonstrates that buffers containing uniform habitats are very rare – only 4 areas have 263 264 such buffers, and they are mostly present at the smallest size (347 m). So even at the scale of our 265 smallest putative range size habitat mosaics are almost ubiquitous. For area C, 10.42% of the 266 smallest buffers were uniform (n=5), for area B it was 7.32% (n = 6), and areas F and G have two 267 uniform buffers each (1.15% and 1.22% of the sample respectively). In total of the 596 buffers 268 analysed at this smallest size, 2.52% (n=15) were uniform (or non-mosaic) habitats. Of these, at 347 269 m one was continuous grassland (<25% canopy cover, area G), one was discontinuous grassland (25-270 50% canopy cover, area F)), seven were closed woodland (>75% canopy cover, area C (n=5), area F 271 (n=1), area G (n=1)), and the remaining six were all semi-desert (area B). For the next largest buffer 272 size, there were only two uniform patches, one of closed woodland in area C and one of semi-desert 273 in area B, while one of closed woodland was still present in area C in the 1199 m buffer.

274 Percentage of canopy cover

275 Table 5 shows the median and range values for the closed woodland (>75% canopy cover) within the 276 different buffer sizes. While area A and area G show a >7% increase in median closed canopy cover 277 from the smallest to the largest buffers, there is relatively little variation in medians within the other 278 images and buffer sizes. Continuous grassland (<25% canopy cover) results are shown in Table 6. 279 Overall the continuous grassland category is less common across images (except area C and area G), 280 and again (with the exception of area G) the medians do not differ greatly across buffer sizes. Tables 281 7 and 8 give the same information for the open woodland (75%-50% canopy cover) and 282 discontinuous grassland (50-25% canopy cover). These also suggest that, with the exception of area 283 G and area A, the median amounts of cover within each category do not change as buffer sizes

increase. In essence, these data are surprisingly scale independent, suggesting that the medians ofthese land cover types tend not change as the estimated home range sizes get larger.

286 We also compared the distribution of fraction of wooded canopy cover (%fwc) calculated from 287 pedogenic carbonates from the Nachukui Formation, Koobi Fora (Quinn et al., 2013) with % of 288 canopy cover observed for the central points of our randomly placed buffers in our seven different 289 study areas (Fig. 3). The Nachukui Formation (2.4-1.4Ma) was chosen as a representative dataset for 290 an East African hominin locality, for which the %fwc data had been published (Quinn et al. 2013). 291 The central point from each buffer yielded a result such as '>75% canopy cover' and those that were 292 not on the forest-grassland continuum (i.e. bare, semi-desert, seasonal water and agriculture) were 293 removed from the analysis. The data generated from these central points are directly analogous to 294 those obtained from pedogenic carbonate analyses, as they are both spatially constrained estimates 295 of canopy cover based on an individual point. However, this does not mean that the results of these 296 analyses are analogous. Our examination of areas from Ethiopia to South Africa (Fig. 1) show that 297 there is great variability in land cover at single points within each image (Fig. 3), and that the 298 expected land cover types are found in the appropriate landscapes. For example, 74 out of 82 points 299 are semi-desert or bare in area B (the Turkana region, data not shown), while area F (encompassing 300 the Kruger National Park) shows considerable heterogeneity with the majority of vegetation 301 recorded as discontinuous grassland or open woodland (25-75% canopy cover). Comparison of these 302 canopy cover distributions with the pedogenic carbonate data (Fig. 3) show that none of our modern 303 images with randomly selected buffers has a similar distribution of habitats to that calculated from 304 the pedogenic carbonates, although the pattern in the Nachukui Formation at Koobi Fora is most 305 similar to our results from the southern side of the Rift valley in modern Ethiopia (area A). However, 306 what is striking is the relative lack of closed habitat reconstructed from the pedogenic carbonates. In 307 the Nachukui formation only 1.45% (n=1) of the total sample are reconstructed with >75% canopy, 308 while in all of our study areas bar that of the Turkana region (area B, 0%), closed woodland 309 comprises 14-45% of the central points examined. We also calculated the %fwc for pedogenic 310 carbonates from three key hominin regions (Busidima Formation, Ethiopia, and the Koobi Fora and 311 Olorgesaile Formations, Kenya) using data from Levin (2013). For the 1475 datapoints only 19 (1.3%) 312 represent closed canopy woodland, suggesting considerable under-representation of these 313 environments in the palaesol datasets.

314

315 Transect results

The results of our 21 transects in areas B and F are best viewed as illustrative of the possibilities in modern Africa, rather than the more quantitative random sampling of buffers described above. They were chosen by eye, to allow us to take a 2.5km transect from a channel, without interference from roads, agriculture, etc. We here present results that illustrate the variety seen in these transects, and the resulting potential for these types of analyses.

321 Patch richness. The greatest patch richness in the majority of transects were found by lakes and 322 rivers, with a decrease in PR as the buffers move away from the water source (Fig. 4a-d). In both 323 area B (Kenya) and area F (South Africa) the changes are seen most abruptly in the smallest 324 estimated hominin home range size (347 m), and vary least with the largest (2500 m). This is 325 expected, as the largest buffer size would still be including the channel and its environs until at least 326 2510 m away from the water source. Transects in area B have lower numbers of habitats in total 327 (Fig. 4a, b), in comparison with area F (Fig. 4c,d), and there is greater complexity in area F (i.e. PR 328 increases and decreases rather than declining as distance to the water source increases).

329 Land cover. Our analyses show the land cover at each 10 m point along a 2500 m transect (Fig. 5) in 330 a way that would approximate to a standard field survey, however with remote sensing the 331 transects can be extended for much greater distances than is often feasible in the field. Figure 5 332 shows two contrasting transects for area B (Turkana). Transect a starts at the edge of Lake Turkana 333 and is largely semi-desert with patches of bare, except for two small patches of open woodland at 334 approximately 2000 m and 2100 m. Transect b starts at the edge of the Turkwel River. The area of 335 semi-desert at the start is an island in the braided channel, and then the main area of riparian 336 woodland is reached and continues for ~620 m, until an abrupt transition to semi-desert and bare 337 patches, with smaller areas of open woodland between 1000-1500 m. The transects for area F (Kruger National Park) show contrasting results (Fig. 5,c,d)- transect c is almost entirely 338 339 discontinuous grassland, with small patches of continuous grassland and a small area of closed 340 woodland at the channel edge. Transect d begins in seasonal freshwater and is largely open 341 woodland throughout its length, interspersed with small patches of closed woodland. Most notable 342 is that at approximately ~1900 m the transect crosses another stream, which is indicated by the 343 increase in closed woodland at this point. Fig. 5 also shows the calculated %fwc for a fossil 344 pedogenic carbonate transect taken lateral to the river in the Dana Aouli Formation, Ethiopia, dated 345 to ~2.7 Ma (Levin et al., 2004) and converted to our land cover classification. This is a much shorter 346 transect (240 m), but lacks any woodland, suggesting an open area, perhaps more similar to that of transect c without the riparian woodland zone. 347

349 Discussion

350 The spatial resolution of palaeoenvironmental reconstructions seldom, if ever, allow an analysis of 351 the range of resources within a hominin's home range. Here we have developed an alternative 352 complementary approach to such questions in hominin behavioural ecology by using modern African 353 vegetation as an analogue to ask questions about the effect of different putative range sizes on 354 access to resources and the potential effects of water sources on such analyses. We have examined 355 what vegetation types the hominins could access at different home range sizes, and concluded that 356 range size does not greatly affect the number of available habitats at these scales of analysis. Even 357 the smallest (perhaps unrealistic) estimated home ranges have habitat variability similar to the 358 largest. In a palaeoecological context this is an encouraging result as it suggests that the fact that we 359 do not know what the range sizes were for early hominins is not crucial if just considering access to a 360 broad range of vegetation types. However, these results are for randomly placed sampling points 361 and range size does make more of a difference if there is a river present (or absent, see below).

Pedogenic carbonates are our current best method of reconstructing the woody cover of past 362 363 habitats. However, they may underestimate the densest cover (in part as a result of preferentially 364 forming in more open soils), and are limited to those regions with soils that are appropriate for their 365 formation (Quade and Levin, 2013). While our land cover results are a 1 year snapshot, the 366 carbonates are time-averaged over (possibly) 100s-1000s of years. Thus the pedogenic carbonates 367 give a less complex 'smoother' result with less woody cover than those landscapes that we see in the modern day. We see analysis of satellite imagery as a complimentary method to that of the 368 369 pedogenic carbonates to examine the potentially 'missing' components within the landscapes. It can 370 be used to match the spatial heterogeneity picked up by fine-scale analysis within individual 371 palaeosols (e.g. at Gona, Quade & Levin (2013)), and allow us to quantify landscape characteristics 372 on continuous regional scales, rather than localised point samples. Remote sensing methodologies 373 can also be used to examine spatial heterogeneity in regions where pedogenic carbonates do not 374 form. While we have not attempted here to directly match modern landscapes to any reconstructed 375 palaeoenvironments, there is clearly the potential to do so. While we cannot match the same 376 vegetation in exactly the same locality as seen in the past, it is likely that a similar mixture of water, grassland and trees can be identified in a number of locations, allowing quantifiable models to be 377 378 built in future.

Our buffers were randomly located in each image, and the results show that patchy ('mosaic')
habitats are very widespread. However, the transect data indicate that the greatest number of
habitats (PR) is found near water courses (Fig. 4), and that these are likely to exert an influence for

382 some distance from the channel (Fig. 5). The habitat variability that water sources introduce are not 383 just the obvious ones of water and riparian woodland, but also for example, seasonal water, swamp, 384 and bare (in gravel bars, etc.). This is important for two reasons - firstly, the highest habitat 385 variability appears to be around water courses, the places that hominins appear to have preferred, 386 but also the places with the appropriate sediments for fossilisation and recovery (see Kullmer, 2007). 387 This has led to suggestions that hominins may not have preferred these habitats, but that we are 388 finding them there as a result of recovery bias (White, 1988). Quinn et al. (2013) demonstrated that 389 lithic sites were significantly more wooded than localities sampled at random within the Koobi Fora 390 and Nachukui formation, and that this association was maintained through the 1.0 Ma time span of 391 deposition. However, the mean woodland density through time was 40 % fwc (Quinn et al. 2013), 392 which in our classification would equate with discontinuous grassland. Either thicker woodland was 393 very rare at the time, in comparison with the present day (Fig. 3), or hominins were seeking these slightly more wooded sites within open areas. Quinn et al. (2013) examine four possible reasons for 394 395 the association of lithics sites with discontinuous grassland, including the need for shade, access to 396 drinking water, raw materials or specific foodstuffs, and explicitly link them to the presence of water 397 (and by association C3 vegetation, shade and cobbles for tool making). In agreement with Quinn et 398 al. (2013), our results suggest that if hominins preferred areas with high habitat heterogeneity (i.e. 399 'mosaic'), then the water courses of Africa are the places to find them, and while we have largely 400 focussed on rivers, the same may also be true of lakes and other water sources (Ashley et al., 2009). 401 This does not rule out the taphonomic explanation but does show that, as mentioned by Roe (1997) 402 and Plummer (2004), there are very many plausible non-taphonomic reasons (beyond just access to 403 drinking water) for utilizing sites near rivers.

404

405 Conclusions

Modern African land cover is highly complex, yet at the scales proposed for hominin range sizes, this may be less important than we may have first thought. A key factor driving the variability in habitats appears to be the presence of rivers (or other water bodies), and this result fits with other proxies used to reconstruct hominin palaeoenvironments. Overall, remote-sensing will not and cannot replace these proxies, but it does allow for examination of spatial complexity on a much large scale than is usual, and may also highlight land cover types and patterns of distributions that many have been present but are under-represented in the fossil record.

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- 423

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- 551

General habitat	Land cover class and code	Description
Woodland	Closed woodland (CDW)	Closed woodland (75%-100% tree cover)
	Open woodland (ODW)	Open woodland (50%-75% tree cover)
Grassland	Continuous grassland (CG)	Continuous grassland (75%-100% grassland)
	Discontinuous grassland (DG)	Discontinuous grassland (50%-75% grassland)
Anthropogenic classes	Agriculture (AG)	Croplands (>50%) Irrigated croplands
	Built-up (BU)	Tree crops Urban areas and settlements Roads
		Quarry and open-cast mine
	Closed coniferous woodland	Coniferous plantation
	(CCW)	Felled coniferous plantation
Bare	Bare (BA)	Bare soil
		Bare rock
		Bare gravel (braided rivers)
		Stony desert
		Sandy desert and dunes
		Salt hardpans
		Lava flows
Freshwater	Permanent freshwater (PF)	Permanent waterbodies
	Seasonal Freshwater (SF)	Seasonal waterbodies
	Swamp (SW)	Swamps and wetland areas
Coastal	Saltwater (ST)	Seas and oceans
	Mangrove (M)	Mangrove forests
	Littoral sediment (LS)	Littoral sediment
		Littoral rock
	Supra-littoral sediment (SLS)	Supra-littoral sediment
		Supra-littoral rock
	Saltmarsh (SM)	Saltmarsh
Semi-desert	Semi-desert (SD)	Semi-desert (bare ground with scattered bushes)
Ice and Snow	Ice and Snow (IS)	Permanent ice and snow Seasonal ice and snow
Sodic lake	Sodic lake (SLA)	Sodic lake

- 554 Table 2. Image locations, acquisition dates and classification accuracies. Note that the timings of dry
- and wet seasons varied each year, with images selected to best represent the variability in
- 556 vegetation levels between wet and dry seasons.

Area	Location	Dry season image	Wet season	Classification
		acquisition data	image acquisition	accuracy
			data	
A	Ethiopia	7 Mar 2002	24 Apr 2002	82.69%
В	Кепуа	15 Oct 2002	5 Feb 2003	91.84%
С	DR Congo / Uganda	9 Jan 2001	25 Nov 2001	86.02%
D	Rwanda / Burundi	17 Aug 2002	13 May 2002	86.89%
E	Malawi / Mozambique	2 Oct 2000	28 Apr 2001	81.64%
F	South Africa / Mozambique	19 Dec 1999	9 Apr 2000	84.73%
G	South Africa	31 July 2001	19 Dec 2000	85.39%

559 Table 3. The five hominin-relevant buffer sizes used in this study.

Radius (m)	Diameter (m)	Area (ha)	Basis	Reference
347	694	38	A. africanus	Antón et al. (2002)
1100	2200	380	Hadza	Raichlen et al. (2014)
1199	2398	452	H. erectus	Antón et al. (2002)
2500	5000	1963.5		This study
13000	26000	53093	Kanjera	Braun et al. (2008)

567 Table 4. Median patch richness (and range) results for the 7 study areas at 5 different buffer sizes.

568 PR has been adjusted to remove all anthropogenically-related land covers (agriculture, coniferous

569 plantations and built up areas).

Buffer	Area A	Area B	Area C	Area D	Area E	Area F	Area G
size (m)	(n=19)	(n=82)	(n=48)	(n=31)	(n=78)	(n=174)	(n=164)
347	4 (2,6)	2 (1,4)	4 (1,6)	4 (3 <i>,</i> 5)	4 (2,6)	4 (1,6)	4 (1,5)
1100	5 (4,6)	3 (1,5)	5 (1,7)	5 (3 <i>,</i> 6)	4 (3,7)	5 (3 <i>,</i> 8)	4 (3,6)
1199	5 (4,6)	3 (2,5)	5 (1,7)	5 (3 <i>,</i> 6)	4 (3,7)	5 (3 <i>,</i> 8)	4 (3,6)
2500	5 (4,7)	3 (3,7)	5 (4,7)	6 (4,7)	4 (4,7)	6 (4,8)	5 (4,8)
13000	6 (5,7)	4 (3,7)	6 (5 <i>,</i> 7)	7 (6,8)	6 (5 <i>,</i> 7)	7 (6,8)	5 (5,8)

571

572	Table 5. Median (and range) of % of closed woodland habitat for the 7 study areas at 5 different
J / L	rable of median (and range) of the oreleved moodal and habitat for the total and a same entre

573 buffer sizes (scaled to be 100% across the 4 closed woodland – continuous grassland land cover

574 categories).

Buffer	Area A	Area B	Area	Area D	Area E	Area F	Area G
size (m)	(n=19)	(n=82)	С	(n=31)	(n=78)	(n=174)	(n=164)
			(n=48)				
347	2.5	0	16.8	28.7	42.5	3.9	7.2
	(0 <i>,</i> 80.5)	(0, 100)	(0,	(0, 97.0)	(0 <i>,</i> 98.8)	(0, 100)	(0, 100)
			100)				
1100	9.0	0	17.2	29.5	46.4	5.0	10.8
	(0.3,	(0, 98.2)	(0.1,	(1.6, 88.3)	(1.9,	(0, 96.4)	(0, 89.0)
	59.9)		100)		94.6)		
1199	9.4	0	17.9	30.0	46.3	4.9	11.7
	(0.3,	(0, 98.8)	(0.2,	(1.9, 87.8)	(2.3,	(0, 96.0)	(0, 88.4)
	61.1)		100)		94.3)		
2500	6.4	0	15.8	28.8	46.5	7.9	15.3
	(0.2,	(0, 77.3)	(0.5,	(4.7 <i>,</i> 78.7)	(2.2,	(0, 92.3)	(0, 79.3)
	65.0)		99.9)		91.8)		
13000	10.2	0.3	12.7	29.9	42.3	8.2	19.5
	(1.6,	(0, 52.1)	(1.1,	(15.1,45.4)	(4.7,	(0.6,	(0.3,
	38.9)		99.5)		86.1)	66.7)	47.7)

575

576

578 Table 6. Median (and range) of % continuous grassland habitat for the 7 study areas at 5 different

- 579 buffer sizes (scaled to be 100% across the 4 closed woodland continuous grassland land cover
- 580 categories).

			1			1	
Buffer	Area A	Area B	Area C	Area D	Area E	Area F	Area G
size (m)	(n=19)	(n=82)	(n=48)	(n=31)	(n=78)	(n=174)	(n=164)
347	1.7	0	36.3	2.9	3.7	3.6	36.5
	(0 <i>,</i> 44.7)	(0,0)	(0, 99.1)	(0, 41.2)	(0, 83.0)	(0, 99.8)	(0, 100)
1100	3.1	0	38.7	4.4	6.3	4.3	41.7
	(0, 32.1)	(0,0)	(0, 79.2)	(0, 29.8)	(0, 70.2)	(0 <i>,</i> 96.4)	(2.3,
							99.7)
1199	3.4	0	38.7	4.3	7.3	4.6	41.5
	(0 <i>,</i> 30.5)	(0,0)	(0, 80.1)	(0, 30.2)	(0 <i>,</i> 69.8)	(0 <i>,</i> 96.8)	(2.2,
							99.6)
2500	7.6	0	38.3	4.8	7.6	5.9	42.6
	(0.4,	(0, 0.2)	(0, 77.7)	(0.6, 25.9)	(0.1,	(0, 92.6)	(5.2,
	35.6)				67.0)		96.1)
13000	9.2	0	42.6	5.2	8.6	8.6	45.9
	(2.0,	(0, 2.1)	(0, 69.4)	(2.4, 17.3)	(1.5,	(0.3,	(18.9,
	27.7)				53.6)	75.1)	92.4)

581

582

583 Table 7. Median (and range) of % of open woodland habitat for the 7 study areas at 5 different buffer

sizes (scaled to be 100% across the 4 closed woodland – continuous grassland land cover

585 categories).

Buffer	Area A	Area B	Area C	Area D	Area E	Area F	Area G
size (m)	(n=19)	(n=82)	(n=48)	(n=31)	(n=78)	(n=174)	(n=164)
347	3.3	100	2.0	30.8	12.9	34.1	6.4
	(0, 39.7)	(0,100)	(0, 17.2)	(1.0,	(0 <i>,</i> 56.5)	(0, 88.8)	(0, 57.6)
				63.5)			
1100	4.1	100	1.8	32.8	14.4	35.3	8.7
	(0.5 <i>,</i>	(0,	(0, 7.5)	(5.0,	(1.7,	(0.2,	(0, 47.9)
	38.1)	100)		61.4)	57.3)	74.9)	
1199	4.6	100	1.8	33.5	14.6	36.0	9.0
	(0.6,	(0,	(0, 7.1)	(5.0,	(1.5,	(0.2,	(0, 46.0)
	38.0)	100)		62.6)	57.2)	74.6)	
2500	5.1	100	2.2	37.0	14.2	35.2	9.9
	(0.5,	(2.2,	(0.04,	(9.8,	(2.9,	(0.1,	(0.01,
	29.3)	100)	5.6)	57.9)	56.2)	70.4)	37.9)
13000	5.7	99.7	2.7	37.1	17.2	34.7	9.7
	(2.5,	(37.9,	(0.2, 4.0)	(23.9,	(5.5,	(3.4,	(0.6,
	15.2)	100)		43.9)	42.1)	55.8)	23.9)

586

- 588 Table 8. Median (and range) of % of discontinuous grassland habitat for the 7 images at 5 different
- 589 buffer sizes (scaled to be 100% across the 4 closed woodland continuous grassland land cover
- 590 categories).

	1	1	1		1	1	1
Buffer	Area A	Area B	Area C	Area D	Area E	Area F	Area G
size	(n=19)	(n=82)	(n=48)	(n=31)	(n=78)	(n=174)	(n=164)
(m)							
347	76.8	0	33.9	28.9	15.2	30.7	21.0
	(13.2,	(0,	(0, 62.5)	(0, 93.1)	(0, 77.8)	(0, 100)	(0, 73.0)
	98.8)	63.9)					
1100	73.6	0	33.4	25.7	16.6	29.5	21.3
	(15.3,	(0,	(0, 58.2)	(4.6,	(0.6,	(0.4,	(0.3,
	98.2)	69.2)		67.2)	69.6)	96.6)	66.1)
1199	73.2	0	33.3	26.6	17.1	28.9	20.9
	(14.3,	(0,	(0, 57.4)	(4.6,	(0.6,	(0.4,	(0.4,
	98.2)	71.5)		66.4)	70.7)	96.0)	65.2)
2500	68.7	0	36.5	23.8	17.5	29.8	22.2
	(11.9,	(0,	(0, 52.3)	(5.2,	(1.6,	(1.7,	(3.5 <i>,</i>
	95.8)	68.8)		63.0)	70.5)	92.6)	57.2)
13000	63.9	0	37.2	27.6	24.6	34.7	23.7
	(44.9,	(0,	(0.4,	(14.4,	(5.0,	(6.6,	(6.8,
	82.1)	13.4)	49.3)	47.8)	44.8)	79.7)	34.0)

- 593 Figure 1. Location of Landsat ETM+ images used as study areas in this analysis, see Table 2 for details
- of site locations.

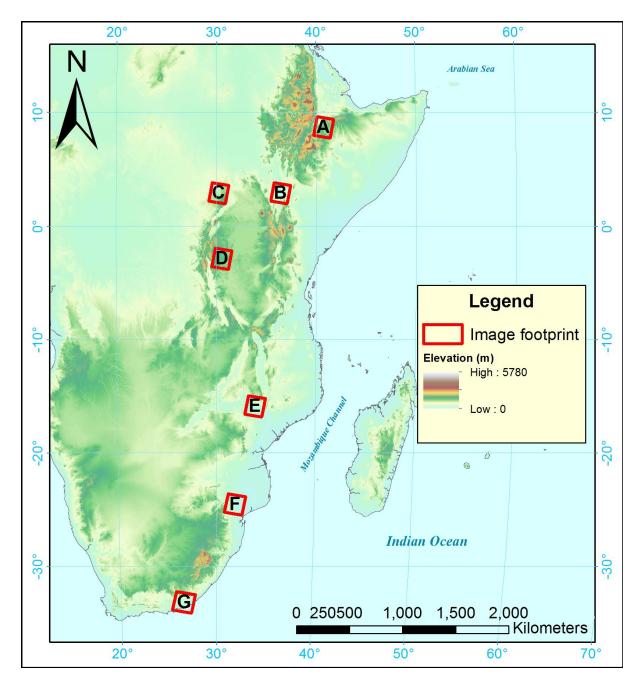


Figure 2. Nested buffers of different radii based on estimated hominin home range sizes, overlaid on
land cover classification for area F, the Kruger National Park, demonstrating the land cover variability
within different radii.

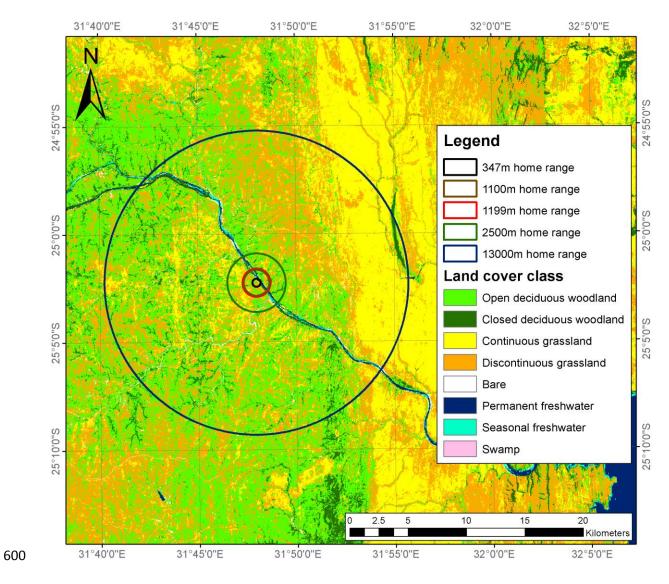
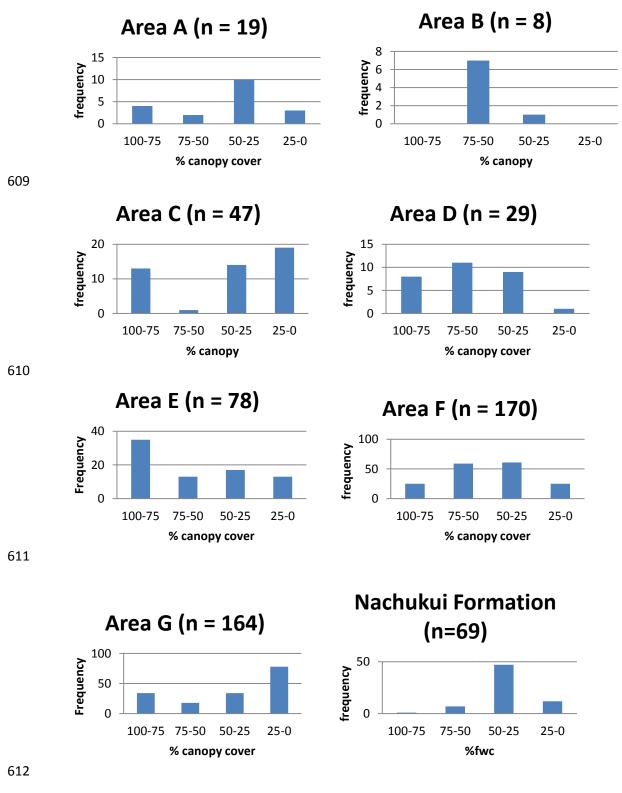


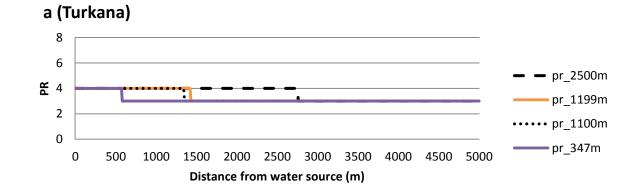




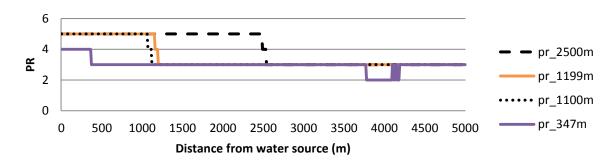
Figure 3. Land cover class frequency for the central points in each randomly placed buffer by study
area (A-G) in comparison with % fraction of woody canopy cover (%fwc) from the Nachukui
Formation, Koobi Fora calculated from pedogenic carbonates (%fwc methods and data from Quinn
et al., 2013). Note: all central points that were not on the forest –grassland continuum (i.e. bare,
semi-desert, seasonal water and agriculture) were removed.



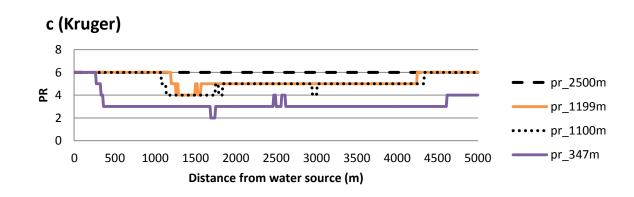
- 614 Figure 4. Patch richness over 5000 m transects in Area B (southern Turkana Basin, Kenya) and Area F
- 615 (Kruger National Park and environs, South Africa). A, b and c illustrate relatively simple transects
- 616 showing PR largely declining with distance moved away from the water source. a) runs south-west
- 617 from Lake Turkana, b) runs south-west from the Turkwel River, c) runs southward from the
- 618 N'wanetzi River, and d) is an example of greater variability in a transect running southward from the
- 619 N'waswitsontso River.

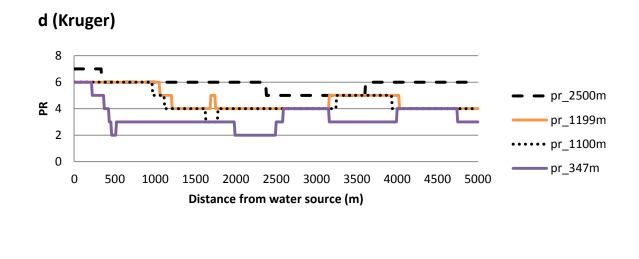


b (Turkana)



621







- 626 Figure 5. Land cover categories recorded every 10 m for 2500 m starting at a water source, for two
- 627 transects from Area B (the Turkana region, Kenya), and two from Area F (the Kruger National Park,
- 628 South Africa). The letters correspond to the PR data shown in Fig. 4 for the same transect locations.
- Also included for comparison is the fraction of woody cover (f_{wc}) calculated from pedogenic
- 630 carbonate data from a palaeosol transect from the Dana Aouli Formation, Gona, Ethiopia (Levin et al.
- 631 2004), and converted to our land cover classifications.

