Movement as a means of social (re)production: Using GIS to measure social integration across urban landscapes

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Movement as a means of social (re)production: Using GIS to measure social integration across urban landscapes

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Abstract
This paper contributes to the archaeological study of movement in urban environments where built forms and natural features worked together to play a key role in structuring human mobility. We propose an analytical method using least cost analysis in a Geographic Information System (GIS) to empirically measure social integration. The method defines mobility as the potential for pedestrian movement, and identifies locations where people were most likely to walk to or through in a landscape. The calculated mobility data are then employed to identify with whom people were most likely to interact and the degree to which they were socially connected with particular groups of society (social networks), and integrated within society as a whole. The results reveal underlying spatial complexities that in conjunction with other archaeological data can be correlated to social, political, or economic inequality in ancient cities. We apply the method to measure social integration between four socioeconomic groups at the Late Classic (AD 600–900) Maya city of Copán, Honduras.

Keywords: Movement, Landscape, Geographic Information Systems (GIS), Least cost analysis (LCA), Ancient Maya, Mobility maps, Social inequality

1. Introduction
Although archaeologists have long considered where ancient people built houses, ritual structures, and civic-ceremonial centers, they have not often considered movement among these features, particularly within cities where built forms and the natural landscape together guided movement. Recent years have seen new archaeological thinking on movement, mobility, and circulation (e.g. AAA, 2010; TOPOI, 2011; Snead et al., 2009; White and Surface-Evans, 2012), including several methods using Geographic Information Systems (GIS) (e.g. Llobera et al., 2011; Taliaferro et al., 2010). The following summarizes key insights underlying many of these approaches, and presents one empirical way to measure social integration in an urban setting.

The spatial organization of settlements and landscapes reflected and shaped ancient life (De Certeau, 1984; Giddens, 1984; Goffman, 1983; Jakobson, 1980; Morgan, 1984; Parmentier, 1987; Peirce, 1966; Silverstein, 1976). People often locate themselves and organize their surroundings to facilitate or restrict access, channel movement, and send visual cues to influence interaction (e.g. Ashmore and Knapp, 1999; Doyle et al., 2012; Hillier and Hanson, 1984; Lawrence and Low, 1990; Llobera, 2000; Rapoport, 1990; Smith, 2011b). Features of the natural landscape, such as topography or hydrology (Anaya Hernandez, 2001b; Rahn, 2005), as well as the human-made built environment, such as houses or streets, affect how individuals and groups move within a place.

We define mobility as the potential for movement across a landscape (Inomata, 2004:179). Consideration of potential, rather than actual, paths of movement allows us to model spatial relations on the scale of neighborhoods, cities, or regions and infer the kind and intensity of social relations between members of society. We use GIS (ESRI ArcGIS 10.0 and 9.3) to compute the cost of movement across a landscape, and derive least cost paths from one location to others. We argue that the average value (i.e. travel time) of these paths is a useful proxy for estimating movement between locations. People are more likely to travel to and interact with people living at places they can more easily, or quickly, reach, because such places are more accessible. Therefore, we assess the social integration of an urban center by calculating how different groups of people were able to move and interact with others within it.
2. Studying social (re)production through mobility

Rodrigue et al. (2009) define accessibility as the measure of the capacity of a location to be reached (attractiveness), or the capacity to reach different locations ( emissiveness). Locations of high accessibility are more easily reached, and are likely to have more people travel to or through them than locations of low accessibility. Such locations are also more connected to, or integrated with, a system as a whole (e.g. a city), or sub-parts of the system (e.g. neighborhoods). Research using space syntax—an approach that measures how the spatial structure of cities influences movement—has established a one-to-one relationship between accessibility and social integration: if a place is easily accessible, it is also highly integrated with the city-wide economy and social life ( Hillier, 1996; Hillier et al., 1993).

Therefore, the amount of movement to, through, or from a place strongly correlates with “integration” in urban systems. Integration values measure how easily a place can reach or be reached from all other locations in the system (Bafna, 2003; Hillier, 1996:160; Ratti, 2004:489). While most space syntax approaches that calculate integration values (particularly in archaeology) are based on axial maps (e.g. Fisher, 2009; Parmington, 2011; Stuardo, 2003), scholars are developing alternative approaches with GIS (e.g. Jiang and Crampton, 2002; Jiang and Gimblett, 2002; Ratti, 2005; Shelton, 2012). For example, the cost-of-passage function calculates the accumulated cost of moving within a street network using an Urban Digital Elevation Model—a raster surface that stores heights of the urban surface (Ratti, 2004, 2005; Ratti and Baker, 2003). Urban landscapes with clustered housing and informal settlement patterns such as sites throughout the Basin of Mexico, Singhalese and Khmer settlements, or Classic Maya cities, tend to be strongly influenced by environmental features and agrarian practices (Fletcher, 2009; Smith, 2010; Stark and Osa, 2007). Therefore, it is necessary to incorporate cost parameters, such as topography, hydrology, and land cover that are not traditionally considered in urban analyses of movement. To account for these factors, we propose a least cost analysis (LCA) approach.

LCA subscribes to Zipf’s Principle of Least Effort (1949), which postulates that people tend to economize their behavior by following paths that take the least amount of energy or time to traverse (White and Surface-Evans, 2012) (Figure 1). In GIS, least cost path analysis identifies the route(s) with the lowest accumulated cost(s) based on user-defined cost parameters. The average value of all least cost paths to or from a location corresponds to what we call an integration value. Locations with higher average pathcosts are (1) more costly to reach, (2) less likely to have people travel to or through them, and (3) less integrated than locations with lower average pathcosts. These average least cost path values serve as a proxy to measure the integration or segregation of groups of people across urban landscapes.

1. See Llobera and Sluckin (2007) for a review of the energy vs. time debate.
King, 1995; Scarborough et al., 2003; Watanabe, 2004). While regional and temporal variation catalyze these debates, archaeologists have had some success correlating architecture to social organization. Typologies based on building and plaza form, composition, size, and complexity relate to the function(s) (e.g. administrative, ritual, domestic) and socioeconomic status (e.g. royal, elite, non-elite, rural) of occupants (Becker, 1971; Kintz, 1983; Tourtellot, 2004). At Palenque, Mexico, as people entered public spaces they encountered imagery promoting dynastic authority, whereas those who entered restricted spaces witnessed the king in a role subordinate to gods, giving us an insider’s perspective to city-level sociopolitical relationships (Parmington, 2011).

At Copán, researchers developed a typology that classifies architectural groups (discrete clusters of individual buildings) into five site types (Fash, 1983; Leventhal, 1979; Willey and Leventhal, 1979). The Harvard Site Typology equates mound size and quantity, and the organizational complexity of structures and associated plazas to socioeconomic status. The types range from non-elite households (types 1 and 2) to elite residences (types 3 and 4) to the royal acropolis (type 5) (Table 1). While only a preliminary model—and one based primarily on physical size—we use the typology as a starting point to select social groups across landscapes (Anaya Hernandez, 2001a, 2001b, 2006; Murrieta-Flores, 2010, 2012; Rahn, 2005). Our work is unique in that we focus on the relationship among spatial organization, mobility, and social inequality at the urban scale. We relate social inequality to social integration, and access to elite or state-sponsored events. Given the close relationship between spatial order and mobility, we expect to identify mobility patterns that reflect Copán’s social hierarchy. That is, we expect that average travel times to and from elite groups (types 3 and 4) to be faster than to and from non-elite groups (types 1 and 2). If these expectations are met, then it can be argued that elites resided in accessible areas.

Table 1. Copán typology (from Willey and Leventhal, 1979, pp. 82–83).

<table>
<thead>
<tr>
<th>Type</th>
<th># of mounds</th>
<th># of plazas</th>
<th>Mound height (m)</th>
<th>Construction</th>
<th>Total # (urban core)</th>
<th>Total # in sample</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>3–5</td>
<td>0–1</td>
<td>0.25–1.25</td>
<td>Earth fill, undressed stone rubble</td>
<td>134</td>
<td>20</td>
</tr>
<tr>
<td>2</td>
<td>6–8</td>
<td>1–2</td>
<td>2.50–3.00</td>
<td>Mostly undressed, but some dressed surface stone</td>
<td>68</td>
<td>13</td>
</tr>
<tr>
<td>3</td>
<td>6–8</td>
<td>1–2</td>
<td>3.00–4.75</td>
<td>Much more dressed stone</td>
<td>20</td>
<td>9</td>
</tr>
<tr>
<td>4</td>
<td>8+</td>
<td>2+</td>
<td>4.75–10</td>
<td>Large stones, rough and dressed, vault stones</td>
<td>13</td>
<td>7</td>
</tr>
</tbody>
</table>

Figure 2. Map of Eastern Mesoamerica. Inset—urban core, Copán, Honduras.
2. To automate the process, we developed a python script (modified from Sherrill et al., 2010). The script runs in ArcGIS 9.3 and 10.

Figure 3. Formulas for calculating integration values, emissiveness and attractiveness.

\[
E \text{ (Emissiveness)} = \frac{\sum o_i + \sum o_2 + \sum o_n}{n} \\
A \text{ (Attractiveness)} = \frac{\sum b_j + \sum b_2 + \sum b_n}{n} \\
I \text{ (Integration value)} = \frac{\sum (o_i + o_2 + o_n) + (b_j + b_2 + b_n)}{2n}
\]

\(o = \text{travel time out (derived from individual least-cost paths from source to destinations)}\)  
\(b = \text{travel time back (derived from individual least-cost paths from destinations to source)}\)  
\(n = \text{number of destinations}\)

locations and/or the accessibility of elite sites increased over time, while the opposite would be true of non-elites. Our expectations do not deny bottom-up community organization and local control of ritual, agricultural, or other activities (Isendahl and Smith, 2013); rather our analysis is of the urban core, and in fact the method could be applied to other analytical scales or used to identify intermediate zones (Arnauld et al., 2012). In terms of broad social (in)equality, large differences in travel cost between different places suggests that settlement across the urban landscape fostered inequality; vice versa, small differences in travel cost between different places might imply greater equality between residents.

4. Methodology

This article presents revisions to a least cost analysis (LCA) method to measure social integration originally proposed by Richards-Rissetto (2010 & 2012). The method measures the potential accessibility of architectural complexes in urban landscapes that exhibit informal settlement patterns using raster data (comprising valued pixels) to calculate accessibility over a contiguous surface rather than a vector-based topological network (e.g. space syntax) (see Richards-Rissetto, 2012 for comparison with space syntax). Using mobility patterns as a proxy measure for potential accessibility, we acquire data on (1) degree of social integration, i.e. are certain groups more or less integrated or segregated from society as a whole, and (2) social connectivity, i.e. who is likely to interact with whom. These data inform understandings of social inequality because they provide information about who might have greater access to social, economic, or political opportunities. Our recent work has modified the original method in two ways: (1) arbitrary units (based on an ordinal scale offering a relative comparison of costs) have been converted to time units (seconds/meter), and (2) both attractiveness and emissiveness are measured (the original measured only emissiveness). Emissiveness and attractiveness provide information on access to resources and opportunities, channeling movement and visual messages that relate to social control and political and economic relations.

The method we present differs from typical least cost path analyses. Most archaeological applications using least cost paths seek to identify actual travel routes or corridors (Bell and Lock, 2000; Kantner, 1997); however, it can be difficult to identify ancient paths due to resolution and quality of topographic data, landscape changes over time, and different least cost computational algorithms depending on the software (but see Harris, 2000; Herzog and Posluschny, 2008). Therefore, our objective is not to identify actual travel routes, but to calculate average travel time to and from specific locations as a proxy for the degree and nature of social interaction. To do this, we calculate integration values using a least cost adjusted-Shimbel index. While the Shimbel index calculates the minimum number of paths necessary to connect one location

with all other locations in a network, the least cost adjusted-Shimbel index calculates the average least cost of paths to or from a source point to all potential destinations in a network (Rodrique et al., 2009). (Figure 3 presents the formulas applied in this study.) Using average least cost also normalizes the data, permitting a comparison of costs between different sub-groups with a different number of source or destination points, a common scenario in archaeology and exemplified below. Another advantage of this method is its utility for comparative analysis across multiple spatial scales, including individual architectural groups, neighborhoods, or whole cities, and between different temporal periods, for example the Early and Late Classic at Copán.

In the case study, we apply this least cost method to four site types from Copán’s urban core to derive quantitative data on mobility. We use these data to generate “mobility maps” that show patterns of potential movements for different groups based on average travel time from one place to another. Mobility maps help us to investigate (1) degrees of social integration for different socioeconomic groups, (2) connectivity between socioeconomic groups, and (3) social (in)equality as defined by differential access to economic, social, or political opportunities. To select a representative sample of type I–4 sites, we employed a stratified random sampling technique. Our sample consists of 49 architectural groups (arguably, households), and represents about 21% of urban core sites (Table 1).

4.1. Procedure Steps² (Figure 4)

4.1.1. Step 1: Friction surface

The first step is to create a friction surface to represent the difficulty (as a percent of total cost) to cross a single cell (pixel). The friction surface incorporates conduits (e.g. roads) and barriers (e.g. buildings or slope greater than x°), combining features of the natural and cultural landscapes. For
the case study, structures, reservoirs, and areas with a slope greater than 40° were assigned as complete barriers. Quebradas (stream cuts) were partial barriers and the cost of movement was weighted by a factor of 1.8, preventing travel within quebradas yet permitting movement across them. While we do not account for seasonal variation (e.g. intermittent quebradas or vegetation cycles) such differences could easily be incorporated into the analysis. We considered Copán’s two sacbeob (causeways) as conduits (e.g. Chase and Chase, 2001). To decrease travel time along them, we weighted the cost of movement by a factor of 0.7. We employed Map Algebra to integrate these weighted variables into a single raster surface, i.e. the friction surface (Figure 5: left).

4.1.2. Step 2: Speed surface
The second step is to create a speed surface that represents how fast one can travel (in seconds per meter) across a single cell. Travel speed is derived from inputting slope (degrees) into a walking algorithm. We generated the slope surface for Copán from the 2 m Digital Terrain Model Richards-Rissetto digitized from the Proyecto Arqueológico Copán I (PAC I) survey maps (Fash and Long, 1983), and inputted it into Tobler’s hiking function (Figure 6) to generate a speed surface in kilometers per hour (Tobler, 1993); however, other algorithms can replace Tobler’s (see Pandolf et al., 1977; Van Leusen, 2000). To decrease travel time along them, we weighted the cost of movement by a factor of 0.7. We employed Map Algebra to integrate these weighted variables into a single raster surface, i.e. the friction surface (Figure 5: left).

4.1.3. Step 3: Travel cost surface
The third step is to generate a travel cost surface that represents total time required to move from one cell to another cell, accounting for conduits and barriers. This surface is generated by multiplying the friction surface (step 1) and the speed surface (step 2) into a single raster surface. Travel cost is a function of three general categories (conduits, barriers, and slope) that researchers can tailor to specific applications. For Copán, travel cost is a function of four factors: (1) sacbeob speed, (2) slope speed, (3) complete barriers (architecture and slope > 40°), and (4) partial barriers (quebradas) (Figure 7). While the travel cost surface incorporates conduits and barriers to movement, we acknowledge that it remains a simplification or abstraction of the real world that could be improved upon in the future. (Natural and cultural factors such as such as terrain type (e.g. bog, sand), land cover (e.g. deciduous, grassland), compelling features, avoidance areas, and other social phenomena (e.g. sacred places, social memories, trade, or alliances) (Munn, 1996; Stanton and Magnoni, 2008) could all be incorporated.)

4.1.4. Step 4a/4b: Minimum accumulated cost surfaces (emissiveness & attractiveness)
The fourth step is to create travel-time surfaces, which represent the minimum accumulated cost to move from a start location to a destination or set of destinations (Llobera et al., 2011; Van Leusen, 2000). In some urban areas topography played a major role in structuring human movement (such as the highlands and southern lowlands of the Maya area); the cost to arrive at a particular place (attractiveness) will not be the same as the cost to leave a particular place (emissiveness). As Copán exemplifies such a situation, we created two time surfaces (travel time out and travel time back) for every source location. The time surfaces use anisotropic modeling to take into account the influence of slope direction (i.e. upslope or downslope) on movement (Kantner, 2004; Wheatley and Gillings, 2002).

The emissiveness and attractiveness surfaces were generated using the PathDistance tool to account for actual surface distance (bumpy vs. flat surface) and for direction dependent costs (i.e. upslope vs. downslope). The inputs were: (1) source site (one of our randomly chosen architectural groups), (2) the travel cost surface, and (3) a linear vertical graph (based on slope). We selected a linear vertical factor, which decreases speed for uphill travel and increases speed for downhill travel; other vertical graphs (e.g. inverse linear) can be used. In total, we generated 98 travel time surfaces: an emissiveness and an attractiveness surface for each of the 49 architectural groups considered in our case study.

3. Studies indicate that maximum slope for pedestrian movement varies based on factors such as age, activity, load, perception, and slope direction (Kinsella-Shaw et al., 1992; Pofflit et al., 1995).
4. Multiple forms exist for converting cost to time traveled, but Tobler’s hiking function has been found to be the most reasonable estimator for travel time in rough terrain (Kantner, 2004:327).
4.1.5. Step 5a/5b: Generate least cost paths – (emissiveness) & attractiveness

The fifth step is to run cost paths, which signify the minimum travel time required to reach or return to a location. Using the emissiveness surfaces created in Step 4, we generated least cost paths from each source site to all destinations (n = 235) based on site type (e.g. source site to type 1 sites, then to type 2 sites, etc.). Using the attractiveness surfaces from Step 4, we generated least cost paths back from each destination based on site type to each source site (see Figure 8 for one illustration of paths and travel times).

Because the least cost paths were generated from the travel-time surfaces, travel time (in seconds) along each path was automatically calculated. The raster paths were converted to shapefiles with attribute tables that stored travel times, and then exported to calculate average travel times for (1) individual architectural groups and (2) site types 1-4.

4.1.6. Step 6a/6b/6c: Calculate emissiveness, attractiveness & integration values

The final step is to export the data (travel times out and back) for each of the source sites, and aggregate them based on significant variables (e.g. site type, neighborhood affiliation, or proximity to particular resources) for the study area under investigation. We generated data sub-sets (as least cost paths in ArcGIS) according to site types 1-4 for each source site. Then we exported these data sub-sets to Microsoft Excel, for a total of 392 tables. Finally, we aggregated the data from these tables according to (6a) emissiveness (travel time out), (6b) attractiveness (travel time back), and (6c) integration values (average of travel time out and back).

6a and b: Emissiveness and Attractiveness: For each source site, we exported four data tables with emissiveness values (converted to minutes; 196 tables), and with attractiveness values (converted to minutes, 196 tables) for a total of 392 tables. The emissiveness tables recorded travel costs from architectural groups (source sites) to destinations classified according to site type, and the attractiveness tables vice versa. Data from these 392 tables were aggregated into eight tables, one per site type (1-4) for each emissiveness and attractiveness, in order to calculate average travel times from all source sites to each of Copán’s four site types.

6c: Integration values: To calculate average travel times to and from site types (1-4), we aggregated the emissiveness and attractiveness data to create a total of four tables.

5. Results

5.1. Summary statistics

Table 2 presents aggregate travel times based on site type. The emissiveness data indicate that travel times from Type 4 (highest-order elite) sites to other sites were the fastest ($t = 12.82$). In contrast, travel from Type 1 (lowest-order non-elite) sites took the longest ($t = 15.56$). Travel times from type 2 and 3 sites were similar ($t = 14.12$ and $t = 14.07$, respectively). The attractiveness data indicate that travel times to Type 4 (highest-order elite) sites to other sites were the fastest ($t = 12.56$). In contrast, travel to Type 1 (lowest-order non-elite) sites took the longest ($t = 15.88$). Travel times to type 2 and 3 sites were similar ($t = 14.34$ and $t = 14.19$, respectively). Averaging emissiveness and attractiveness times to generate integration values indicates that travel times to and from type 4 (highest-order elite) sites to other sites were the fastest ($t = 12.70$). In contrast, travel to and from type 1 (lowest-order non-elite) sites took the longest ($t = 15.72$). Travel times to type 2 and 3 sites were similar ($t = 14.23$ and $t = 14.14$, respectively). The emissiveness and attractiveness data as well as the integration values corroborate each other.

5.2. Mobility patterns

From the summary statistics, we generated mobility maps to compare the mobility patterns of Copán’s four socioeconomic groups, which show a correspondence between mobility and socioeconomic status. Generally, the mobility patterns reflect social order as hypothesized by the Harvard Site Typology (i.e. type 4 sites at most accessible locations and type 1 sites at least accessible locations). However, contrary to our expectations, type 3 and 4 “elite” sites did not have similar travel times, nor did type 1 and 2 “non-elite” sites. Instead, travel times for type 2 and type 3 sites were almost identical (Figure 9). These results confirm Richards-Rissetto’s earlier work (2010), suggesting that the distinction between type 2 and 3 sites should be reevaluated. Subsequent test excavations at type 2 sites recovered “elite” architecture and artifacts at presumed non-elite sites (Landau, 2013; Webster et al., 2000). These archaeological findings suggest that mobility patterns indeed reflect and influence social structure, perhaps more than architectural group size or organizational complexity. The least cost approach aligns well with excavation results, and both undermine a one-to-one correlation between size, complexity, and status for ancient Maya architecture and planning.

6. Discussion

6.1. Social integration

At Copán, integration values from the urban core indicate that residents at non-elite type 1 sites were the most segregated group, and residents at elite type 4 sites were the most integrated group (Table 2; Figure 9). The low social status of residents occupying type 1 sites was underscored by their spatial placement in more segregated areas. While residents of type 1 sites may have had easier access to agricultural fields or local shrines, they had to spend more time to witness major ritual events, pay tribute at the civic-ceremonial core or elite sites, or acquire seasonal water rations from the city’s

5. For this paper we did not perform statistical analyses, but packages such as MiniTab and SAS may be used (see Richards-Rissetto, 2010).

6. Interestingly, one surveyor commented that the on-the-ground distinction between type 2 and 3 sites was the least clear of all (Leventhal, 1979:43).
reservoirs (Davis-Salazar, 2003; Fash, 2005). Conversely, the elites who resided at type 4 sites could more easily attend events at the civic-ceremonial core or routinely interact with other urban core dwellers, particularly Copán’s rulers, whose royal temples and residence they could more quickly reach (emissiveness) (Richards-Rissetto, 2010, 2012). The placement of elite complexes at accessible, highly integrated locations suggests that to attend state-sponsored events or carry out

![Figure 7. Travel cost surface—represents total time to move from one cell to another cell (includes barriers/conduits).](image1)

![Figure 8. Emissiveness map (left) showing cost paths from Group 9N-8 (an example start site) out, and attractiveness map (right) showing cost paths from destinations back to Group 9N-8.](image2)
economic exchanges in the civic-ceremonial group, lower status people moved past elite households adorned with ornate sculpture atop high platforms that conspicuously displayed a message of high status, wealth, and prestige (attractiveness) (Dahlin et al., 2007; Shaw, 2012; Trigger, 1990). Figure 10 illustrates how travel to type 1 sites requires passing by type 4 sites, while the opposite is not necessarily true. Through daily routines and bodily movement the elite may have created the means to forge and reinforce economic ties, social cohesion, and political support (Rapoport, 1990; Smith, 2007).

6.2. Social connectivity

While the integration values indicate that type 1 residents lived at segregated locations and type 4 residents lived at integrated locations, we also analyze the disaggregated (emissiveness and attractiveness) data to investigate which socioeconomic groups were more likely to interact with other groups of the same type. Table 3 lists round-trip times between specific site types. These data offer insight into intragroup and intergroup relationships, by indicating a pattern of hierarchically structured relationships—elite living at type 4 sites were able to establish strong social connections with members from their own socioeconomic class, while simultaneously segregating members of lower classes from each other. Travel between type 4 sites was, on average, seven minutes faster than travel from elite type 4 sites to non-elite type 1 sites, indicating that residents of type 4 elite sites were more strongly connected to members of their own social class then a lower class. In contrast, intragroup travel between lower status residents took much longer; travel between type 1 sites took almost six minutes longer than travel from type 1 to type 4 sites. For Copán’s 3 km² urban core, this difference equates to ca. 25% longer travel times that may reflect intentional intragroup segregation and channeling of lower status residents to or past type 4 complexes to establish and reinforce social groups on an economic and/or political basis.

### 7. Conclusions: Movement as a means of social re(production)

Movement is a cultural mechanism (or practice) that structures the flow of information and enables interactions that produce and reproduce social networks (Bourdieu, 1977; Giddens, 1984; Murrieta-Flores, 2010; Rapoport, 1990). Because mobility patterns reflect the potential for face-to-face interaction, they can be read as social networks, and thus provide information on social integration, social connectivity, and finally, social inequality. This paper presents a method to empirically measure social integration based on mobility patterns.

Within Copán’s urban core, our preliminary results show that the potential for movement across the landscape correlates with social status. Our least cost mobility analysis revealed relationships between spatial and social inequalities that help refine prior models of socioeconomic hierarchy, especially in regard to “middle-level” type 2 and 3 settlement. On one hand, significant difference in mobility between type 4 and 1 sites indicates a hierarchical socioeconomic structure; on the other hand, the lack of difference between type 2 and 3 sites seems to reveal internal variation—possibly masked by hierarchical thinking—and points to presence of more fluid or heterarchical categories. In other words, consideration of mobility among architectural groups suggests heterarchical relationships within an overall hierarchy, as has been proposed for other sites in the Maya area (e.g. Potter and King, 1995; Tourtellot et al., 2003).

For the ancient Maya, mobility analysis reveals a more complex picture of social organization than traditional size-ranked site typologies. Perhaps consideration of movement across the built and natural landscape—physical structures and the “empty” space between them—will be useful for understanding the social and political organization of ancient Maya cities, a topic where no significant agreement exists (Chase and Chase, 2004; Watanabe, 2004). Future work will statistically evaluate the mobility data, and analyze and compare features at selected architectural groups to generate more refined hypotheses for mov...
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subsequent archaeological investigation. As well, further investigation of physical and/or cultural factors that influence movement for the ancient Maya specifically, or any cultural group generally (e.g. Fisher, 2009), will add much to the accurate calculation of mobility and social integration.

Our methodology can be applied to social categories other than socioeconomic status at Copán (e.g. ethnicity, age, gender) or to other cities characterized by variable topography and informal street networks (Smith, 2007, 2011a). Quantitative measurement will facilitate comparative analysis between cities, and may offer empirical strength to theories of state formation and change that consider social integration as a determining factor (e.g. Blanton and Fargher, 2008; Earle, 1997). Because the spatial layout of urban landscapes influences daily movement and shapes personal interaction, mobility studies help to illuminate social structures and how they may be produced and reproduced.

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**References**


**Table 3.** Round trip (travel time) for Copán’s urban core sites.

<table>
<thead>
<tr>
<th>Start</th>
<th>Destination</th>
<th>Round trip (min)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Type 4</td>
<td>Type 4</td>
<td>22.24</td>
</tr>
<tr>
<td></td>
<td>Type 3</td>
<td>23.40</td>
</tr>
<tr>
<td></td>
<td>Type 2</td>
<td>27.01</td>
</tr>
<tr>
<td></td>
<td><strong>Type 1</strong></td>
<td><strong>28.89</strong></td>
</tr>
<tr>
<td>Type 3</td>
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<td><strong>33.93</strong></td>
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</table>

**Figure 10.** Comparison of travel between Type 1 and Type 4 sites in Copán’s urban core.
Anaya Hernandez, A., 2006. Strategic location and territorial integrity: The role of subsidiary sites in the Classic Maya Kingdoms of the Upper Usumacinta region. Internet Archaeology 19.


