



Remotely identifying potential vector habitat in areas of refugee and displaced person populations due to the Syrian civil war

Samuel N. Chambers,¹ Joseph A. Tabor²

¹*School of Geography & Development;* ²*School of Natural Resources and the Environment, University of Arizona, Tucson, AZ, USA*

Abstract

Historically leishmaniasis is most prevalent in established urban centres but this research shows that refugees and, most significantly, internally displaced persons are now commonly in areas characterized by the presence of fly habitats potentially leading to higher prominence of *Leishmania* infection. Areas engulfed by the Syrian civil war has thus caused the dispersal of humans into previously unpopulated areas amid habitats of the sand fly *Phlebotomus papatasi* that hosts the parasite *Leishmania*. The addition of new places of exposure to this disease add to difficulties with respect to diagnosis as well as provision of care and treatment. We used geospatial methodology adapting it to remotely identifying and analyzing sand fly habitats with the aim of measuring how common it is. Our methodology helps avoid the issue of resolution in satellite imagery by measuring likelihood rather than strictly known locations. We followed up this information with spatial analysis identifying which civilian populations are most prone to sand fly exposure, and therefore leishmaniasis, due to their geographical situation. Our results suggest that those most

likely to be exposed to *Leishmania* are internally displaced persons, those camps less likely to receive medical relief and typically having temporary residents migrating elsewhere.

Introduction

Leishmaniasis, a parasitic protozoan disease spread by *Phlebotomus papatasi* sand flies is a major cause of infections and skin ulcers in times of war, both with regard to military staff (Cross *et al.*, 1996) and to civilians (Hayani *et al.*, 2015; Inci *et al.*, 2015; Al-Salem *et al.*, 2016; Özkeklikçi *et al.*, 2017) The World Health Organization (WHO) reports that the disease is responsible for 20 to 30 thousand deaths in the world each year, with an estimated 900 thousand to 1.3 million cases annually (WHO, 2016). Historically, leishmaniasis has been documented as most prevalent among the impoverished in urban areas, often also common among military personnel, in what is now northern Syria and south-eastern Turkey (Schilling, 1921). The risk for *Leishmania* infection is a growing concern with the continuation of the Syrian Civil War and the associated refugee crisis (Hayani *et al.*, 2015; Inci *et al.*, 2015; Al-Salem *et al.*, 2016; Özkeklikçi *et al.*, 2017). An estimated 4.8 million Syrians have been displaced externally (Response, 2016). Refugee sites are located within areas with a relatively high probability of a high density of sand fly habitats, particularly in Lebanon, Kurdish Iraq and south-eastern Turkey, which have a long history of recorded cases of leishmaniasis (Pigott *et al.*, 2014), while reports from Kurdish fighters suggest this as a relatively new phenomena in the rural areas (Rudaw, 2015). Although this new development has been locally attributed to the disposal of corpses (Rudaw, 2015; Mondragon-Shem and Acosta-Serrano, 2016), the high prevalence of leishmaniasis is likely the result of the absence of disease control programmes and disruption of health services due to the war (Mondragon-Shem and Acosta-Serrano, 2016).

Cross *et al.* (1996) developed a method based on remote sensing to predict the distribution of *P. papatasi* sand flies in the Middle East based on the Normalized Difference Vegetation Index (NDVI). These authors showed that the highest risk of exposure and spread of the disease was highest from May to October coinciding with NDVI values of 0 to 0.06 (Cross *et al.*, 1996). The use of NDVI and land cover classification are well suited as low-cost rapid risk assessment of vector-borne illnesses (Brownstein, 2004). Our study expanded on this methodology in combination with a geospatial analysis of the likelihood of sand fly habitats near sites of refugees and displaced persons in Syria and bordering countries. People trapped within areas inaccessible to health care would theoretically be of particular concern if they were also located in an area with a high likelihood of exposure to the parasite. Such populations would be difficult to detect by health care

Correspondence: Samuel N. Chambers, School of Geography & Development, University of Arizona, 1064 East Lowell Street, Tucson, AZ, USA.

Tel.: +1.520.465.1721.

E-mail: schambers@email.arizona.edu

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providers increasing the risk for local outbreaks (Berkley, 2017). To identify such risks geographically in war-torn areas, methods would need to be rapid and, unlike traditional methods, rely almost entirely on remote assessment rather than sampling on-site vector populations (Diuk-Wasser *et al.*, 2006). As shown by Brownstein *et al.* (2002), rapid assessment of vector-borne illness requires remote sensing and predicative models.

Leishmaniasis was typically confined to urban areas such as Aleppo, Syria (Yaman and Özbel, 2004) but cases have spread with occurrences among refugees (Al-Salem *et al.*, 2016). One of the biggest limitations to an analysis of the current leishmaniasis status in refugee and displaced populations is the difficulty gathering accurate information due to the Syrian Civil War (Du *et al.*, 2016). This analysis seeks to circumvent that problem by relying on place and likelihood of habitat rather than documented records, which are difficult to collect and maintain during war and mass migration. The question becomes a question of exposure to infected sand flies. This study seeks to answer that with applied remote sensing and a statistical analysis of known populations, and their locations, impacted by the Syrian civil war.

Materials and Methods

Within our study area, with a centroid location of 35.65N and 39.71 E, we developed a remote sensing model based on Moderate Resolution Imaging Spectroradiometer (MODIS) data acquired by the satellites Terra and Aqua belonging to the Earth Observing System developed by the US National Aeronautics and Space Administration (NASA). The eMODIS data collection (Hansen *et al.*, 2004) provided the normalized difference vegetation index (NDVI) based on visible and near-infrared bands of the electromagnetic spectrum which is capable of measuring the presence of live green vegetation in a 250×250 m resolution (Rouse *et al.*, 1973). MODIS NDVI has a range of values from -1 to +1 by the formula $(NDVI-4000)/6000$. All values from 0 to 0.06 coincided with sand fly habitat locations as shown in previous analyses (Cross *et al.*, 1996). We converted these continuous spatial data to point locations with a 250 meter Kernel density, *i.e.* the density of points surrounding each point (Bithell, 1990; Silverman, 2018), to identify the areas of highest probable density of habitats (PDH). The Kernel density gives a smoothed map where we can infer habitat density based on the spacing of detected locations. The neighbourhood radius is standard to kernel density by geospatial analysis and determined by the formula (Eq. 1):

$$R = 0.9 * \min (SD, \sqrt{\frac{1}{\ln(2)} * Dm}) * n^{-0.2} \quad \text{Eq. 1}$$

where SD is the standard distance, Dm the median distance, and n the number of points. Standard distance was calculated as (Eq. 2):

$$SD = \sqrt{\left(\frac{\sum_{i=1}^n (x_i - \bar{X})^2}{n}\right) + \left(\frac{\sum_{i=1}^n (y_i - \bar{Y})^2}{n}\right)} \quad \text{Eq. 2}$$

where x_i and y_i are the coordinates for the point, with \bar{x} and \bar{y} representing

the mean centre, and n the number of points (Mitchell, 2005). We did this for the month of June 2015 since this month represents the peak season for sand fly dispersal in the region (Cross *et al.*, 1996; Orshan *et al.*, 2010). Areas of zero kernel density values, were considered as having a near zero or lower probability of containing sand fly habitats at the time. PDH values were extracted by point locations for known refugee locations (mostly official United Nations camps), known as Internally Displaced Persons (IDP) camps, and the major cities of Syria for those persons still confined there.

Results

Every refugee camp location was found to be situated within a sand fly habitat. A visual inspection of the model (Figure 1) suggests that the camps near Antakya, Turkey were especially at risk with high PDH and relatively near past cases of leishmaniasis according to data collected by Pigott *et al.* (2014).

In 2014, there were an estimated 6.8 million IDPs within Syria (Sharara and Kanj, 2014). The majority of these were shown to be in locations near the Turkey and Lebanon borders or outside cities. However, the two IDP camps on the south-eastern border with Jordan showed a PDH of zero. The details of leishmaniasis cases is lacking for IDPs except for some unverified information from Kurdish soldiers (Rudaw, 2015). What is evident from the PDH is that the IDPs are at high risk, but unlike the refugee camps, the reality on the ground is less clear (Figure 2).

An analysis of the major Syrian cities shows that all but one were situated within low to high PDH areas. Like the IDP camps, the details on the ground are less certain (Figure 3).

A single-factor ANOVA (Sokal and Rohlf, 1969) shows IDP camps as the overall highest average PDH values with the major cities having the lowest. However, with their wide distribution across the borders, the refugee sites did have the highest variance. The P-value is less than the α of 0.05, so the Refugee Sites, IDP camps, and the cities are significantly different with regard to their PDH values. A post-hoc Least Significant Difference (LSD) anal-

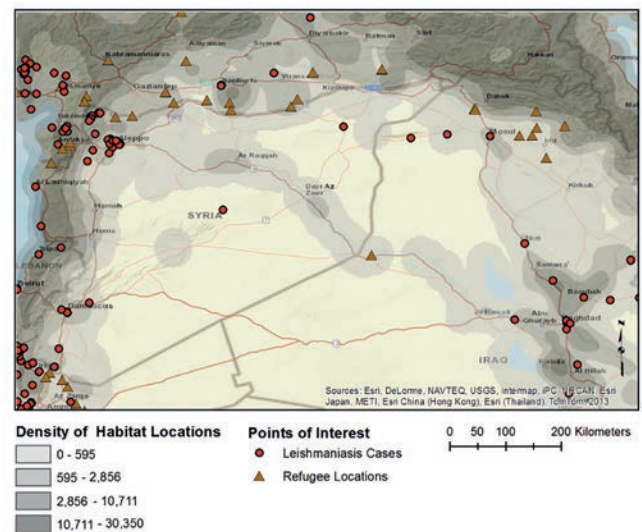


Figure 1. Probable density of habitats (in points per km²) in relation to refugee locations and historically known points of occurrence of leishmaniasis (data derived from Pigott *et al.*, 2014).

ysis of the ANOVA values showed each site group being significantly different from all other group. This shows that, those persons overwhelmingly most likely to be exposed to *Leishmania* are those displaced in camps within Syria.

Boxplots show a large fourth quarter for refugee sites with an even greater outlier (Altinozu 1, Hatay province, Turkey) and a larger range than the IDP camps. Cities also show outliers (Tartus and Latakia, Syria) reaching as high as the third quartile of IDP camps, but with a much smaller median that is below the first quartile of the IDP camps. The maximum PDH values for IDP camps were in the Khirbat al Jawz IDP Camps 1 & 2 (Idlib Governorate, Syria) (Table 1).

Outliers include Tartus (127), Latakia (123), and Altinozu 1 (30) (Figure 4).

Histograms suggest these outliers and differing ranges as resulting from the bimodal distribution of each location variable. Each variable has sites within high and low PDH values but IDP camps have a larger number of records with moderate PDH values,

while Refugee sites have the overall highest PDH values but also the highest amount of low PDH values, which demonstrates the importance of individual site condition in planning and observation (Figure 5).

Discussion

With the conflict and movement of people, leishmaniasis has begun to be documented within refugee camps (Al-Salem *et al.*, 2016). Our results support this including the sudden jump in new cases in Lebanon from 6 in the 2000-2012 period to 1,033 in 2013, 96% of which were among Syrian refugees (Alawieh *et al.*, 2014). There is also the issue of new *Leishmania* species introduced into areas with large population migration (Koltas *et al.*, 2014). All of this suggests that a combination of the following factors: i) lack of health care; ii) likelihood of sand fly habitats and; iii) mass

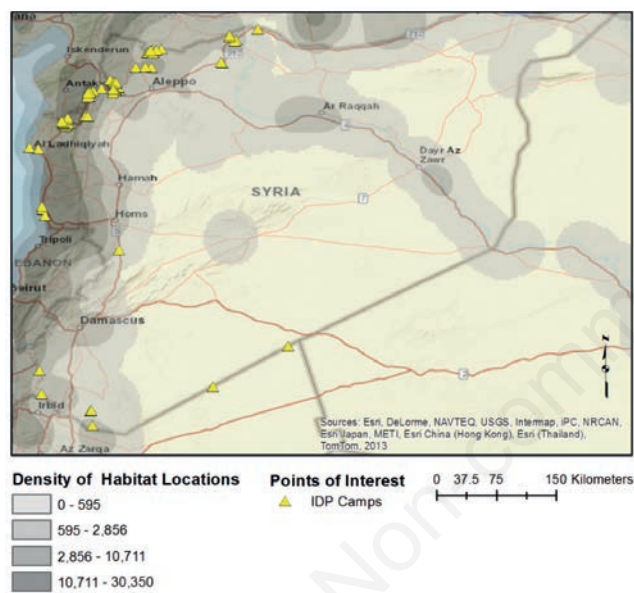


Figure 2. Probable density of habitats (in points per km²) in relation to internally displaced persons (IDP) camps (data derived from Pigott *et al.*, 2014).

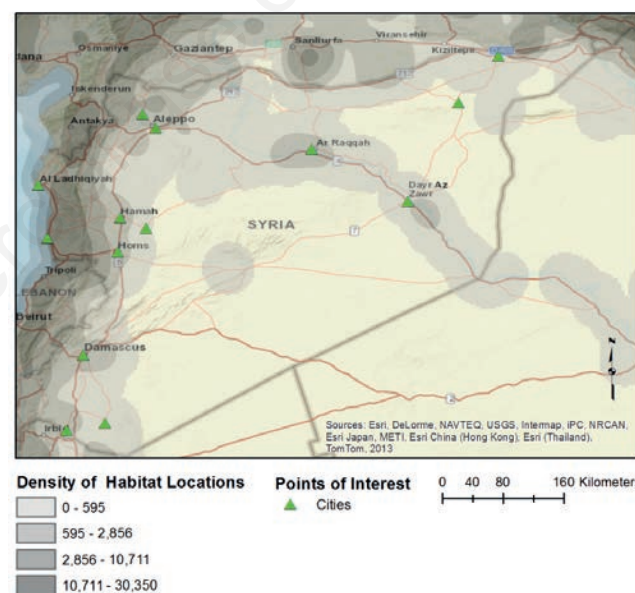


Figure 3. Probable density of habitats (in points per km²) in relation to Syrian cities (data derived from Pigott *et al.*, 2014).

Table 1. Summary of probable density of habitats values for refugee sites, internally displaced persons camps (IDP), and Syrian cities.

ANOVA Source of variation	SS ^a	df ^b	MS ^c	F ^d
Between groups	203416385	2	101708193	4
Within groups	3463191830	128	27056186	-
Total	3666608215	130	-	-
	Refugees & IDP	Refugees & cities	IDP & cities	
Least Significant Difference	1,863	2,976	2,796	-
Difference	35	27	62	-

^aSum of squares (SS) is the squared differences of each observation from the total mean; ^bDegrees of freedom (df) are the count of values that are free to vary; ^cMean square (MS) is the number of df of the sample variance; ^dF-statistic (F) is the variation between the sample means.

migration of persons contribute to the increase of *Leishmania* cases in refugees. The latter factor has two outcomes: settlement of refugees in camps and spread through migration, which ultimately means that the routes and sites matter when planning for the health care and understanding future threats.

The rapid risk assessment methodology we developed provides a spatial epidemiological use for planning relief in the case of displaced and trapped persons in conflict. Our combined use of remote sensing and kernel density calculation allows not only identifying the places likely to have sand fly habitat but also provides a measure of the likelihood of such habitats regionally. This PDH measurement can serve as a proxy for concluding risk or potential exposure to *Leishmania* when on-site analysis is not feasible. We find this crucial as there are narrow time windows (Chen *et al.*, 2008; Zidi *et al.*, 2013) for treating persons in conflict zones. This is especially the case within camps, where people have left their homes and are temporarily settled in another area, and it is of paramount importance due to the lack of immunity to regional strains of *Leishmania* following migration of displaced populations (Tripathi *et al.*, 2007; Salam *et al.*, 2014; Du *et al.*, 2016). Although remote in origin, our results provide a picture of higher

resolution of the risks inherent when dealing with vector-borne disease in displacements leading to collapsed health care systems (Du *et al.*, 2016).

IDP camps had the overall highest average PDH values while the major cities showed the lowest. Refugee sites in surrounding nations had the highest variance. The Refugee Sites, IDP camps, and cities had significantly different PDH values. Results show that people displaced within Syria have a higher risk of *Leishmania* exposure compared to the other populations in the study area.

We found it evident that the issue of leishmaniasis in camps should be a concern in Syria and all countries bordering Syria including the Turkish Hatay province, a prominent exiting point from Aleppo, Syria and a known hotspot for sand flies (Yaman and Özbel, 2004). A study comparing January 2009 and July 2015 found that the Syrian Civil War impacted the number of infections in the Southeast of Turkey (Özkeklikçi *et al.*, 2017.) This is especially an issue in the urban areas of the region (Nimer, 2018).

Overall more concern needs to be given to IDP camps as these locations are more difficult to reach with emergency health care (Johnson, 2013) and house temporary residents frequently on the move (Taleb *et al.*, 2015), but have the highest average PDH. Although it was suggested as an issue by Du *et al.* (2016), this was not evident in the analysis by Al-Salem *et al.* (2016) of leishmaniasis and conflict in Syria because only refugee camps were analyzed and there were no predictive spatial methods applied as it was reliant on recorded incidents. Prior studies of other IDP populations have found significant vector-borne illness cases in the war zones of Pakistan (Hussain *et al.*, 2018) and the Democratic Republic of the Congo (Brooks *et al.*, 2017). If possible, these and future locations should be monitored or continue to be monitored for the disease and the presence of sand flies.

Shortcomings of this analysis include verification and scale. As many of these places are difficult to reach, it will be more difficult to verify the presence of sand flies and leishmaniasis cases in different locations, especially those in ongoing war zones such as the IDP camps and cities. The movement of people between locations would also increase this difficulty. There is also the issue of MODIS' cell size of 250 m which leaves room for error due to undetected or falsely identified PDHs. The use of kernel density can partially circumvent this by measuring probability but may miss particularly isolated habitat sites. Another shortcoming is the magnitude of population migration and inability to survey in account for this (Du *et al.*, 2016).

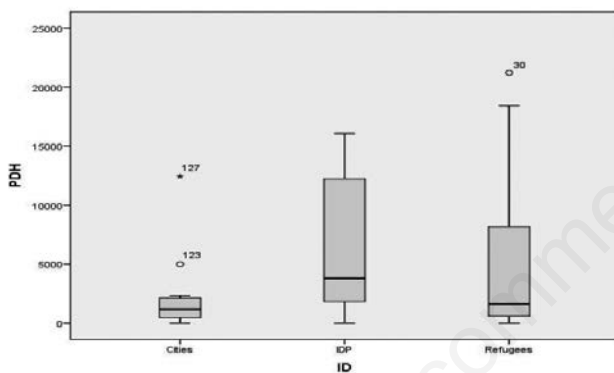


Figure 4. Boxplot of probable density of habitats (PDH) for Cities, internally displaced persons (IDP) camps, and Refugee locations.

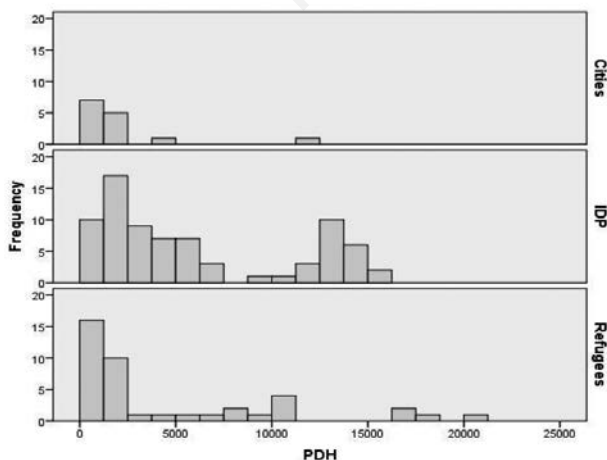


Figure 5. PDH for cities, IDP camps, and refugee locations.

Conclusions

If locations cannot be directly monitored, we can at least know that they have a high likelihood of exposing persons to *Leishmania* infection because of their location. Relief efforts must be a high priority, based on the results of the analysis and specific to site.

Our process and results demonstrate the values of remote sensing and spatial analysis in the planning of relief in conflict zones. It provides a novel geospatial methodology for a continued analysis of PDHs and the potential for leishmaniasis cases in the area. It also provides a baseline for understanding public health in times of conflict, when verified on-the-ground data is in short supply or in difficult to reach locations with short time frames for relief and analysis.



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