

Spatio-temporal Spillover Risk of Yellow Fever in Brazil

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Data Collection

Monthly confirmed cases of yellow fever for each Brazilian Município (sub-state administrative units) from 2001 to 2013 were downloaded from the Brazilian government's portal da saúde website, **tabnet** (<http://tabnet.datasus.gov.br>) on 20 June 2018. Confirmed cases were reported by the Ministry of Health Notification of Injury Information System (SINAN-Net) as determined using clinical-epidemiological criteria.

The annual population for each Município from 2001 to 2013 was also downloaded from the Brazilian government's portal da saúde website, **tabnet** on 05 June 2017. The estimated population was calculated by the Instituto Brasileiro de Geografia e Estatística (Brazilian Institute of Geography and Statistics) as intercensorial estimates.

Monthly land surface temperature and normalized difference vegetation index (NDVI) data from 2001 through 2013 were downloaded from the NASA Land Processes Distributed Active Archive Center (LP-DAAC). The MODIS MOD11C3 product contains monthly temperature data at a 0.05°resolution. THE MODIS MOD13A3 product contains monthly NDVI data at a 1 km resolution. Both gridded data products were then aggregated to the municipality level to obtain a monthly spatially averaged temperature and NDVI value for each municipality.

Rainfall data was obtained from the NASA GESDISC data archive in the form of data from the Tropical Rainfall Monitoring Mission from 2001 through 2013. The 3B43 product contains an average rainfall rate for each month at a 0.25°resolution. We aggregated the gridded data to the municipality level by taking the spatial mean.

Monthly fire locations were downloaded from the Fire Information for Resource Management System (FIRMS). The MODIS Active Fire Product (MCD14ML) reports fire at a 1 km resolution by month. This data was then aggregated to the total number of fires per municipality by month, and scaled to density by dividing by the municipality's total area.

Primate species richness data was obtained from the IUCN Redlist of Terrestrial Mammals for species in genera known to be susceptible to yellow fever (*Ateles*, *Aotus*, *Alouatta*, *Saimiri*, *Cebus*, *Callicebus*, *Callithrix*, *Saguinus*, *Lagothrix*) [1, 2]. Individual species' ranges were combined to calculate the number of species found within a municipality, defined as species richness. We also estimated the relative proportion of primate habitat overlapping with agricultural land use per municipality per year. Shapefiles of geographic ranges of each genus were constructed from the above range maps. Yearly land cover data from 2001 - 2013 was downloaded from the NASA Land Processes Distributed Active Archive Center (LPDAAC). The MODIS MCD12Q1 dataset contains yearly land cover categories at a 1 km resolution by year. The proportion of total municipality area that was both agricultural land use and within a genus range was then calculated for

each genus. These proportions were then summed over all nine genera, resulting in a value from 0 - 9 per municipality by year, defined as the Agricultural and Primate Overlap.

Probability of mosquito vector species occurrence was downloaded from the VectorMap Data portal (<http://vectormap.si.edu/>) on May 30 2018. These are published MaxEnt models of the probability of species occurrences based on environmental covariates and known species presence points. We downloaded data for South and Central America coverage of *Hg. leucocelaenus*, *Hg. janthinomys*, and *Sa. chloropterus* (the only species available that are known sylvatic reservoirs of YF). These models were created in 2011 by the Walter Reed Biosystematics Unit (WRAIR, Division of Entomology). The data is static and at a spatial resolution of 0.04167 °. To calculate the maximum probability of a vector occurrence, we created a spatial average value of vector occurrence for each municipality and species and selected the maximum value of the three species to use for each municipality.

Results

Summary of YF cases in Brazil

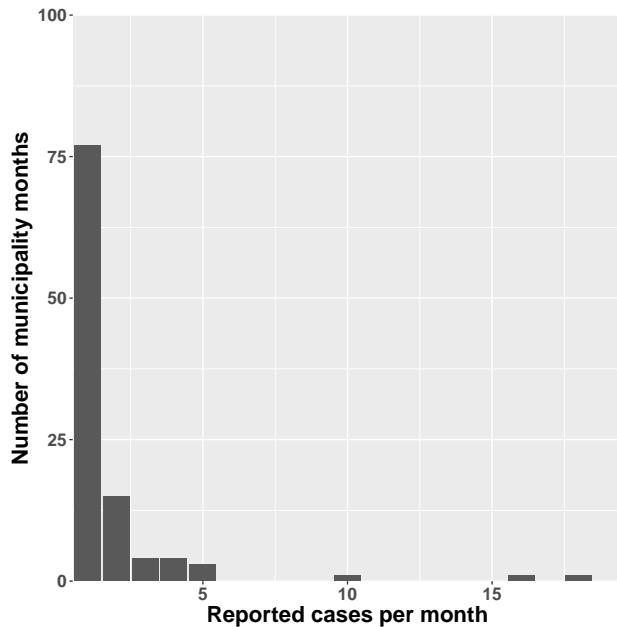


Figure 1:

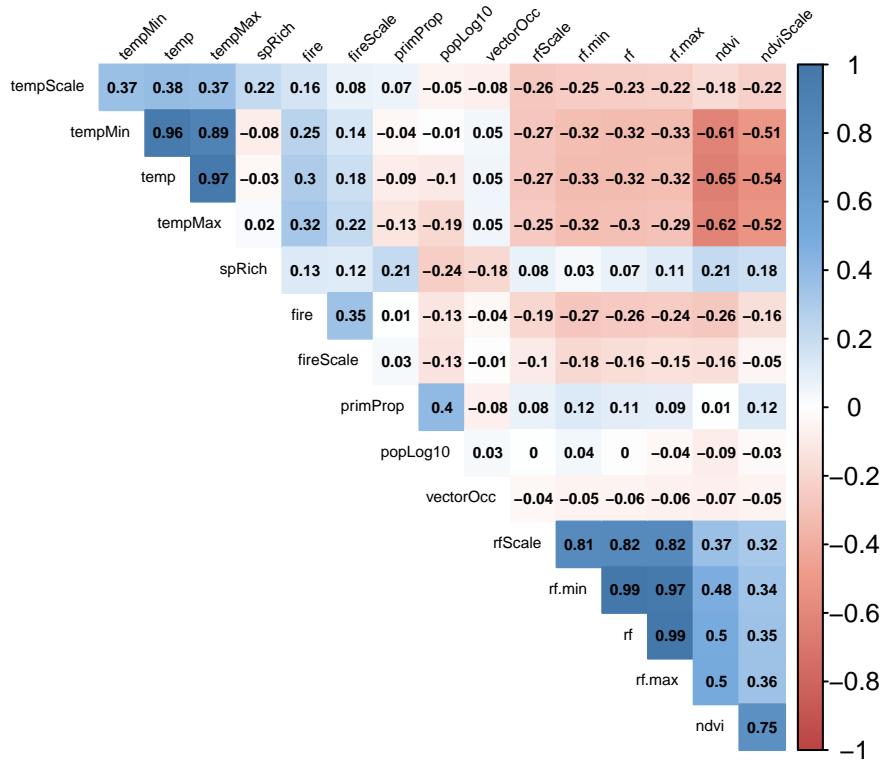


Figure 2: Correlation matrix of full set of covariates. Numeric inset represents Pearson correlation coefficient. Highly correlated covariates were not included in the final model.

Univariate analysis of covariates

Preliminary data extraction and exploration involved the calculation of the spatial minima, mean, and maxima of temperature and rainfall for each municipality-month. However, these values were extremely correlated with each other (Figure 2). Therefore, we conducted univariate analyses of each variable on the national training dataset using a bagged logistic regression with 50 bags to determine if the spatial mean was an appropriate metric to represent spatial variation in these variables and if it had high explanatory power. When comparing AUC between models (our chosen metric of model performance), the ability to predict on the training dataset was comparable amongst types of spatially aggregated variables (Table 1).

Table 1: Results of univariate analyses. Mean AUC values (\pm s.d.) of univariate bagged logistic regression models on training dataset using spatial minima, mean, or maxima of rainfall and temperature.

	Rainfall	Temperature
Minimum	0.631 ± 0.056	0.530 ± 0.55
Mean	0.643 ± 0.093	0.501 ± 0.016
Maximum	0.694 ± 0	0.517 ± 0.038

Long-term trends in YF spillover from 2001 – 2013

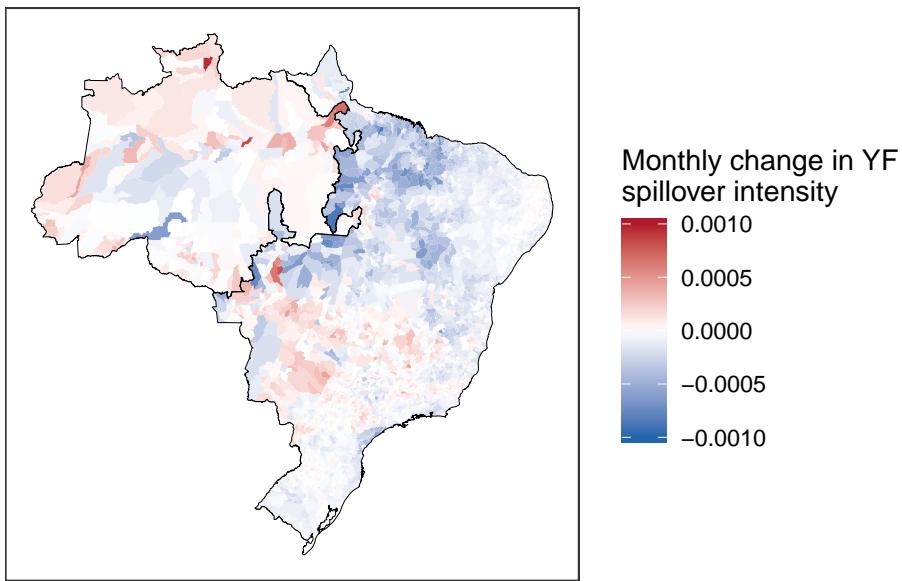


Figure 3: Long-term trends in predicted YF spillover. Results of slope tests exploring the change in predicted spillover intensity in each municipality across all 156 months. Values represent average change over all 156 months. Non-significant ($\alpha = 0.05$) slopes are reported as zero.

Long-term trends in YF intensity by calendar month

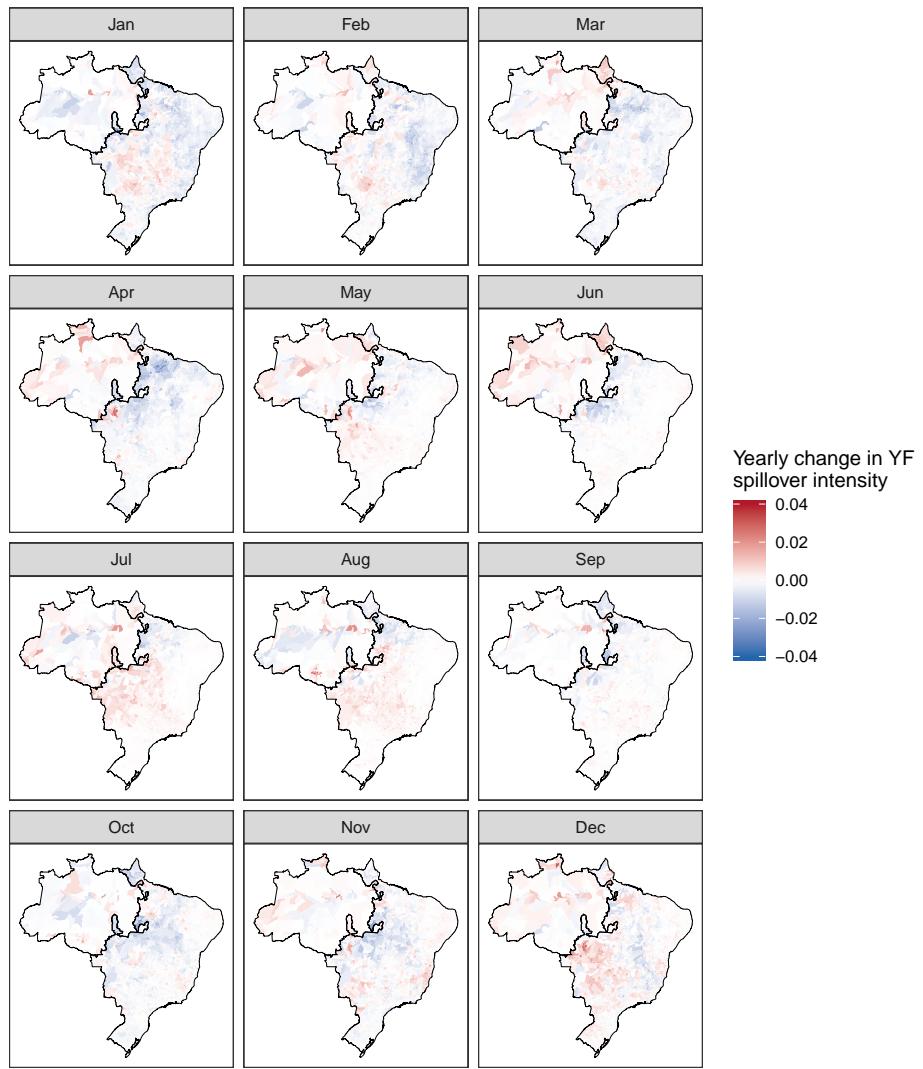


Figure 4: Long-term trends in predicted YF spillover by calendar month. Results of slope tests exploring the change in predicted spillover intensity in each municipality and month of the year from 2001 - 2013. Values represent average yearly change for each month. Non-significant ($\alpha = 0.05$) slopes are reported as zero.

Supplemental results figures

References

- [1] Bicca-Marques, J.C., de Freitas, D.S.: The Role of Monkeys, Mosquitoes, and Humans in the Occurrence of a Yellow Fever Outbreak in a Fragmented Landscape in South Brazil: Protecting Howler Monkeys is a Matter of Public Health. *Tropical Conservation Science* **3**(1), 78–89 (2010). doi:10.1177/194008291000300107
- [2] Hamrick, P.N., Aldighieri, S., Machado, G., Leonel, D.G., Vilca, L.M., Uriona, S., Schneider, M.C.: Geographic patterns and environmental factors associated with human yellow fever presence in the Americas. *PLOS Neglected Tropical Diseases* **11**(9), 0005897 (2017). doi:10.1371/journal.pntd.0005897