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Environmental conditions do not predict diversification rates in the Bantu languages

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ABSTRACT

Keywords:
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Isolation by distance

The global distribution of language diversity mirrors that of several variables related to ecosystem productivity. It has been argued that this is driven by the size of social networks, which tend to be larger in harsher climates to ensure food security, leading to reduced language divergence. Is this pattern purely synchronic, or is there also a quantifiable relationship between environmental conditions and language diversification over time? We used a spatio-temporal phylogeny of the Bantu language family to estimate local diversification rates at the times and locations of language divergence. We compared these data against spatially-explicit reconstructions of several palaeoclimatic and palaeovegetation variables (mean annual temperature and the temperature of the coldest and warmest quarter, annual precipitation and the precipitation of the wettest and driest quarter, growing degree days, the length of the growing season, and net primary production), to investigate a potential link between local environmental factors and diversification rates in the Bantu languages. A regression analysis does not suggest a statistically significant relationship between climatic or ecological variables and linguistic diversification over time. We find a strong positive correlation between pairwise linguistic and geographic distances in the Bantu languages, arguing for a dominant role of isolation as a result of the rapid Bantu expansion that might have overwhelmed any potential influence of local environmental factors.

1. Introduction

The distribution of present-day global language diversity has been shown to correlate strikingly well with several ecologically relevant variables, including climate and altitude (Bentz et al., 2018), biodiversity (Gorenflo et al., 2012; Harmon, 1996; Loh and Harmon, 2005; Moore et al., 2002; Nettle and Romaine, 2000; Sutherland, 2003), the distance from the equator (Bentz et al., 2018; Breton, 1991; Currie and Mace, 2012, 2009; Mace and Pagel, 1995; Nichols, 1999, 1990), and the length of the growing season (Nettle, 1998). A possible causal mechanism underlying these patterns was proposed by Nettle (1998, 1996), who explained a strong correlation between country-level language diversity and the length of the growing season in terms of the environmental risk hypothesis. The environmental risk hypothesis states that intra- or inter-annually fluctuating food supply requires the formation of larger social networks in order to mitigate these variations by means of trading and sharing resources when needed, and thus ensure a continual availability of food. An increasing uncertainty of a stable supply of subsistence products (‘ecological risk’) is argued to require increasingly larger networks, and the social and economic interactions within those require a common linguistic base. The synchronic studies mentioned above rely on spatial correlations between present-day language diversity and environmental variables. Can environmental conditions also explain the diversification of languages over time in a diachronic approach?

Tracking the diversification of languages through time and space requires temporally and spatially explicit language phylogenies. Reconstructions of such phylogenies have been proposed for the Austronesian languages (Gray et al., 2009), the Indo-European languages (Bouckaert et al., 2012), the Tupi languages (Walker et al., 2012) and the Bantu languages (Grellier et al., 2015). Among these, the diversification of the Bantu languages is arguably best understood and mapped (Diamond and Bellwood, 2003), and so we focus on this language family...
here. The expansion of the Austronesian languages is complicated by sea barriers, the location of nodes in the phylogeography of the Indo-European languages are largely uncertain, and the methods used to estimate the timing and location of divergences in the Tupi languages are likely too simplistic.

Grollemund et al. (2015) previously showed that environmental conditions significantly affected the routes and speed of the spatial expansion of Bantu-speaking peoples. Here, we test whether climate and vegetation also affected the rate of diversification of the Bantu language family through time. We use a dated phylogeny to calculate rates of language diversification at the nodes of the phylogenetic tree, and link these to the environmental conditions that were present at the times and locations provided by the phylogeography. Reconstructed environmental variables include mean annual temperature and the temperature of the coldest and warmest quarter, annual precipitation and the precipitation of the wettest and driest quarter, growing degree days, the length of the growing season, and net primary production, thus covering a range of possible drivers behind the ecological risk hypothesis.

2. Materials and methods

2.1. Spatio-temporal language phylogeny and diversification rates

The Bantu languages spread as a result of the expansion south and east of farmers in tropical West Central Africa beginning around 5,000 years ago, and eventually covered most parts of subequatorial Africa, inducing a large-scale cultural shift. The reconstruction and understanding of the Bantu expansion has greatly benefited from the synthesis of linguistic data with genetic, archaeological and biogeographical evidence (Bostoen et al., 2015; Diamond and Bellwood, 2003).Following a spatially-explicit but undated phylogeny (Currie et al., 2012), Grollemund et al. (2015) derived a fully dated tree for 424 Bantu languages using Markov chain Monte Carlo methods and archaeological data to first confine the age ranges of specific nodes before applying a variable-rates molecular clock model to estimate node ages. Based on the geographical locations of extant languages, node-specific latitude and longitude were inferred using a variable-movement rate Brownian motion (see Fig. 1).

We calculate the local rate of language diversification at each node of the phylogenetic tree as follows. Letting \( n_1, \ldots, n_N \) and \( a_1, \ldots, a_N < 0 \) denote the tree’s internal nodes and their ages, respectively. The clade originating at some node \( n_i \) defines a lineage-through-time (LTT) plot, mapping the time interval \([a_i, 0]\) onto the logarithm of the number of ancestral lineages of that clade (Ricklefs, 2007). The latter equals 0 at time \( a_i \) and increases towards the present, where it is equal to the logarithm of the total number of extant languages descending from node \( n_i \). The local diversification rate at \( n_i \) corresponds to the right derivative of the LTT plot at time \( a_i \). Since the LTT plot is given by a discontinuous step function, we define the node-specific diversification rate by means of the slope of a linear regression passing through the point \((a_i, 0)\) of the LTT plot in a suitable interval \([a_i, a_i + \epsilon]\) (Fig. 2). The value of \( \epsilon = 100 \) years was found to be robust, and is used in the analyses.

2.2. Reconstructions of environmental conditions

Environmental variables (listed in Fig. 3) are based on a recent dataset of high-resolution Holocene climate and vegetation reconstructions (Beyer et al., 2019) that is based on downscaled and bias-corrected outputs of the HadCM3 and HadAM3H global climate models (Valdes et al., 2017), and simulations of the Biome4 vegetation model (Kaplan et al., 2003). HadCM3 was previously used to simulate the last glacial cycle in multiple snapshots (see Singarayer and Burrough (2015), for a detailed description of the model setup). We linearly interpolated the available monthly climate data from their 0.5° spatial resolution and 1,000 year temporal resolution to the geographic location and time of each node. In addition to mean annual temperature and mean temperature of the warmest and coldest quarter, annual precipitation and precipitation of the wettest and driest quarter, and net primary production, we also included growing degree days, and the mean length of the growing season in our analysis. Growing degree days (Prentice et al., 1992) were derived using estimates of daily temperature values, which we approximated from the monthly data by means of a cubic spline. The mean length of the growing season, which Nettle (1998) previously linked to global linguistic diversity, is calculated as the number of months when mean temperature lies above 6 °C and is smaller than 0.5 times its total precipitation (in mm).

3. Results

Fig. 3 plots diversification rates of the Bantu family against the environmental conditions present at the time and location of the relevant node of the phylogeny. We do not observe a distinct pattern; linear regressions are weak (\( R^2 < 0.01 \)) and not significant (\( p > 0.07 \)) for all covariates.

These results suggest that if environmental conditions did at all affect the diversification of the Bantu languages, then they did so to a much lesser extent than other factors not considered in our analysis. In addition to, for example, population size (Bromham et al., 2015; Wichmann et al., 2007; Wichmann and Holman, 2009) and sociocultural variables (Gavin et al., 2013) (neither of which we included in our analysis due to lack of data for the relevant time frame), a likely candidate for a more dominant driver of diversification is isolation by distance, given the considerable speed at which the Bantu people spread into distant regions over a very large area, continually resulting in the loss of contact between groups and the collapse of existing social ties. Indeed, we observe a strong relationship between pairwise language divergence and geographical
4. Discussion

Here, we did not identify a significant effect of environmental conditions on the diversification rates of the Bantu languages. This may potentially be due to the rapid nature of the expansion, which could have overwhelmed any possible influence of local environmental factors. Languages that have not undergone such a fast expansion as the one observed for Bantu family might be more informative on the role of the local environment, although suitable spatio-temporal phylogenies on such languages have not yet been derived. A better understanding of the relationship between local environments on linguistic diversity – in combination with existing theory on language evolution (e.g. Perc, 2012; Petersen et al., 2012) – could enhance our understanding of past human migrations and spatial demographics, where archaeological records alone do not provide sufficient information to draw a clear picture. In these cases, the incorporation and synthesis of linguistic and biogeographical data can help to provide additional evidence, test hypotheses and thus complete the picture (Bostoen et al., 2015).

Declarations

Author contribution statement

Robert Beyer, Andrea Manica: Conceived and designed the analysis; Analyzed and interpreted the data; Wrote the paper.

Joy S. Singarayer: Analyzed and interpreted the data; Contributed analysis tools or data; Wrote the paper.

Jay T. Stock: Analyzed and interpreted the data; Contributed analysis tools or data; Wrote the paper

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Competing Interest Statement

The authors declare no conflict of interest.

Additional Information

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