**The importance of language-internal and -external factors during different stages of change**

**A corpus-based analysis of adjective amplification in Hong Kong, Indian, and Philippine English**

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**Abstract**

This corpus-based study examines potential language variation and change in the adjective amplifier systems of Hong Kong, Indian, and Philippine English based on the respective components of the International Corpus of English. The aim of the study is to ascertain if these Asian English varieties align with the global trend according to which innovative variants (*really* or *so*) are replacing *very* as the dominant adjective amplifier in informal spoken English. Furthermore, the study analyses which factors govern the use of adjective amplifiers during different stages of change. The results show that the adjective amplifier systems of Hong Kong and Indian English are remarkably stable across apparent-time while Philippine English shows notable signs of ongoing change. Conditional inference trees that were applied to the data sets confirm that during stability and initial stages of change, as represented by the amplifier systems of Hong Kong and Indian English, intra-linguistic factors figure most prominently while extra-linguistic, social variables gain importance once a change is nearing mid-range, as seen in Philippine English. The results are interpreted to show that

**Keywords**

Intensification, adjective amplifiers, language variation and change, conditional inference tree, very, really, so, Hong Kong English, Indian English, Philippine English

**Introduction**

This study investigates the adjective amplifier systems of Hong Kong English (HKE), Indian English (IndE), and Philippine English (PhiE) the effects of language-internal and language-external factors during ongoing change and evaluates if the importance of these factors differs across stages of change. To address this issue, the present study focuses on ongoing change in the adjective amplifier systems of Hong Kong, Indian, and Philippine English based on data from the International Corpus of English. As adjective amplification has been found to be a site of “fevered invention and competition that would be hard to come by elsewhere” (Bolinger 1972: 18), this domain lends itself to fine-grained analyses of change.

While there exists a substantial amount of research on research on the factors that impact language change (Labov 1994, 2001, 2010) and changes in adjective amplification (e.g. D’Arcy 2015, Tagliamonte 2008, Tagliamonte & Denis 2014), issues relating to when which factors are more or less impactful during changes in this domain remain unresolved. A key factor for why such issues were difficult to address is methodological: whereas most analyses of ongoing change use regression modelling to evaluate the impact of different factors during language change, tree-based models (see Tagliamonte & Baayen 2012) they are more appropriate to detect differences in the impact of factors across stages of change. Therefore, the present study uses Conditional Inference Trees to determine which factors are dominating during which phase of change.

One of the most consistent findings in previous research on changes in adjective amplification in informal spoken discourse has been the replacement of very by really (D’Arcy 2015, Tagliamonte 2008). The present study investigates if this trend holds true for Hong Kong, Indian, and Philippine English and if the underlying factors that drive this change vary across different stages of change in these Asian English varieties.

The present analysis shows that *very* is being replaced by *really* in Philippine English which is similar to the findings from research on change in adjective amplification in inner circle varieties. As such, the change that can be observed in Philippine English represents a vigorous to mid-range stage of change while the change is only in its incipient stage in Indian English. In Hong Kong English, the amplifier system represents a pre-change situation where the amplifier system is remarkably stable. The statistical analysis reveals that during stasis and incipient stages of change, amplifier sue is determined by language internal factors while extra-linguistic factors, in particular social factor, drive change once the change is accelerating and has become more vigorous.

In review of the converging results from previous research and the present study of Asian varieties of English, the paper adds credence to the hypothesis that linguistic change represents an “opportunistic process that reinforces social distinctions by associating them with particular linguistic variants” (Labov 2002). Applied to the present case this means that change is initially dominated by intra-linguistic factors and accelerates once the innovative variant is sufficiently associated with a social group that then serves as the extra-linguistic driving force during vigorous and mid-range change.

**Previous Research**

RQ1: Are Asian Englishes in-line with the trend among traditional L1 and other post-colonial varieties of English in that *very* is being replaced by *really*?

We hypothesise that HKE, IndE, PhiE align with other varieties: *really* is replacing *very* as the default adjective amplifier

In addition, we assume that PhiE which is more closely aligned with AmE and AusE will exhibit a more advanced stage of change while HKE and IndE, which are more aligned with GBE, will be lagging behind due to the more conservative nature of standard BrE with respect to adjective amplification.

**Data and Methodology**

The analysis uses data from the Hong Kong, Indian, and Philippine components of the International Corpus of English (ICE). While the ICE data were collected between 1990 and 2005 and the ICE data is thus not fully up-to-date, they have a shared design which allow us to draw meaningful comparisons between varieties of English. A further advantage of using ICE data stem from the fact that the corpus data is accompanied by socio-demographic information about the speakers which render the ICE components a very useful and interesting resource for variationist analyses. All ICE components consist of one million words (600,000 spoken and 400,000 written) from diverse spoken and written text types. For the current study, only data representing transcribed informal private spoken conversations were considered. This conversational section comprises roughly 200,000 words within each of the components.

The data processing, visualization, and analysis was performed in R (R Core Team 2020). During data processing, the data from the three ICE components was collapsed while retaining information of the origin of each data point. Collapsing the data guaranteed that the same processing steps were applied to the data irrespective of origin while retaining the ability to split the data at a later stage to enable comparisons between the three regional components. The spoken private dialogue section of each component was selected and part–of-speech tagged using the OpenNLP package (Hornik 2016).

Next, all adjectives (pos–tag JJ) were retrieved and it was determined whether adjective were preceded by an amplifier. Identifying amplifiers was based on a list of predefined amplifiers[[1]](#footnote-1) (see footnote XXX) and additional manual inspection. In addition, the sociodemographic information for each speaker that had uttered an adjective was added. This information consisted of the age (16-25, 26-41, 42+), gender (female versus male), and ethnicity of speakers (Chinese versus non-Chinese) as well as the date of the recording (before and including 1998 versus after 1998).

Furthermore, a sentiment analysis was applied to all adjective types (Jockers 2017). The sentiment analysis used in the current study is based on the Word-Emotion Association Lexicon (Mohammad & Turney 2013; cf. http://www.purl.org/net/NRCemotionlexicon) which comprises 10,170 terms and in which lexical elements are assigned scores based on ratings gathered through the crowd-sourced Amazon Mechanical Turk service. Raters were asked whether, according to their judgement, a given word was associated with one of eight basic emotions (ANGER, ANTICIPATION, DISGUST, FEAR, JOY, SADNESS, SURPRISE, TRUST; see Plutchik 1980, 1994). The resulting associations between terms and emotions are based on 38,726 ratings from 2,216 raters who answered a sequence of questions for each word which were then fed into the emotion association rating (cf. Mohammad & Turney 2013). Each term was rated 5 times. For 85 percent of words, at least 4 raters provided identical ratings. For instance, the words dark or tragic are more readily associated with SADNESS while words such as happy or beautiful are indicative of JOY and words like cruel or outraged may indicate ANGER. When applying the sentiment analysis to the present data, an adjective receives a negative-polarity score for every word that is associated with ANGER, DISGUST, FEAR, or SADNESS. For every word associated with ANTICIPATION, JOY, SURPRISE, or TRUST the tweet receives a score for positive polarity. This way, the sentiment analysis allows to test if the emotionality (positive, neutral, or negative) is correlated with changes in the use of specific amplifier variants.

The next step consisted in determining if the same amplifier type had occurred within a span of three adjective slots previously which is necessary to test for potential priming effects Szmrecsanyi 2005, 2006; Tulving & Schacter 1990: 301). Also, the token frequency of adjective type by corpus and age group was calculated to prevent frequency changes in adjective types to confound the analysis (see also Tagliamonte and Roberts 2005).

Next, misspellings of adjectives were corrected using the qdap package (Rinker 2020). Further, elements that were erroneously pos-tagged as adjectives, misclassified items, negated adjectives, and adjectives preceded by downtoners, adjectives that were not amplified by at least two different amplifier variants (to remove lexicalized expressions such as *right honorable*), forms that behave a-prototypical (e.g. *much*, *many*), as well as comparative and superlative forms were removed from the data.

Then, the syntactic context of each adjective was determined (attributive *the nice man* versuspredicative *The man is nice.*). Also, each adjective was assigned a gradability score. This score represents the likelihood of an adjective type to occur in comparative contexts against the likelihood of its occurrence in a non-comparative context. As such, the gradability score reflects the tendency of adjective to be more or less gradable. Also, each adjective was categorized semantically based on Dixon’s (1977) basic semantic adjective types (cf. also D’Arcy (2015); Tagliamonte and Roberts (2005); Tagliamonte (2006, 2008)). However, after testing the usefulness and validity of this fine-grained classification using conditional inference trees, the categories proposed by Dixon (1977) were simplified to only differentiate between adjective that denote value-related meanings (e.g. *good* or *bad*) and all other adjectives. Furthermore, due to the overall sparsity of variants other than *very*, *really*, and *so*, all other variants were collapsed into a bin category *other*. Similarly, all adjective types other than *different* were collapsed into a bin category as the multivariate analysis revealed that *different* was the only adjective type that significantly affected amplifier choice. The final step of data processing consisted in a manual cross–evaluation of the data.

**Statistical analysis**

The statistical analysis of the data uses frequency tabulation to tap into apparent-time rajectories of amplifier use and conditional inference trees (see Hothorn et al. 2006; Hothorn, Hornik & Zeileis 2006) based on the partykit package (Hothorn & Zeileis 2015) to determine which factors correlate with the use of different variants. Conditional inference trees iteratively partition the data so as to minimize residual deviance while maximizing the distinctness of the resulting groups. Appropriate alternatives such as multinomial mixed-effects regression models could also be used but their output is less readily interpretable.

Conditional inference trees were applied to the combined data as well as to each variety separately. This allowed us to determine which factors correlated with amplifier choice overall but also within each variety.

**Results**

The following presents the results of the multivariate analysis which used conditional inference trees to ascertain which factors impact amplifier choice across and with HKE, IndE, and PhiE. We begin with the depiction of the apparent-time trajectory of amplifier types by syntactic context (see Figure 1).

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Figure 1 shows that the amplifier systems of HKE and IndE are remarkably stable according to the ICE data.

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| Figure 1: Conditional Inference Tree for HKE, IndE, and PhiE. |

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| Figure 2: Conditional Inference tree for the ICE Philippine data. |

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| Figure 3: Conditional Inference tree for the ICE India data. |

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| Figure 4: Conditional Inference tree for the ICE India data. |

**Discussion**

Summary

The analysis shows that . . .

- the trajectory of change in the adj. amplifier system of PhiE is approximating the trajectories of other varieties (NZE, IrE, AusE, etc.)(see e.g. D’Arcy 2015)

! only PhiE confirms H11

- change in PhiE is substantially further advanced compared to HKE and IndE ! confirmation of H12

- during stasis and initial stages of change, intra-linguistic factors dominate

- once change is accelerating, extra-linguistic (social) factors become more important (see Labov 2002)

Discussion

- During initial stages of change, the type of the amplified adjective and its linguistic properties determine amplifier use

- Once an amplifier variant attains social meaning, it will (further) increase in use: linguistic change piggybacks on social stratification

“Sociolinguistic variation is parasitic upon such linguistic variation. It is an opportunistic process that reinforces social distinctions by associating them with particular linguistic variants. [. . . ] The use of linguistic forms to increase distinctiveness of particular groups is a driving force for the acceleration of change” (Labov 2002)

Outlook

Could this be a universal mechanism/pattern?

Test if the mechanisms. . .

- can be shown to have worked in analogous changes in

English (e.g. 3rd p. sg. ind. morpheme: <eth> ! <(e)s> )

- can be shown to have worked in analogous changes in languages other than English

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1. absolutely, actually, aggressively, amazingly, appallingly, awfully, badly, bloody, certainly, clearly, dead, completely, considerably, crazy, decidedly, definitely, distinctly, dreadfully, enormously, entirely, especially, exactly, exceedingly, exceptionally, excruciatingly, extraordinarily, extremely, fiercely, firmly, frightfully, fucking, fully, genuinely, greatly, grossly, heavily, highly, hopelessly, horrendously, hugely, immediately, immensely, incredibly, infinitely, intensely, irrevocably, mad, mega, mighty, most, much, obviously, openly, overwhelmingly, particularly, perfectly, plenty, positively, pretty, profoundly, purely, real, really, remarkably, seriously, shocking, significantly, so, specially, specifically, strikingly, strongly, substantially, super, surely, terribly, terrifically, total, totally, traditionally, true, truly, ultra, utterly, very, viciously, wholly, wicked, wildly [↑](#footnote-ref-1)