Cognitive influences in language evolution: Dutch data

Introduction

This is the model code for Monaghan & Roberts, "Cognitive influences in language evolution: Psycholinguistic predictors of loan word borrowing". It takes data from the WOLD database of borrowing for Dutch and tries to predict whether a word has been borrowed or not according to various psycholinguitic measures.

Load libraries

```
library(mgcv)
library(sjPlot)
library(lattice)
library(ggplot2)
library(gplots)
library(dplyr)
library(party)
library(lmtest)
library(gridExtra)
library(itsadug)
library(car)
library(caret)
library(scales)
logit2per = function(X){
  return(exp(X)/(1+exp(X)))
rescaleGam = function(px, n, xvar, xlab=""){
  y = logit2per(px[[n]]$fit)
  x = px[[n]]$x *attr(xvar, "scaled:scale") + attr(xvar, "scaled:center")
  se.upper = logit2per(px[[n]]$fit+px[[n]]$se)
  se.lower = logit2per(px[[n]]$fit-px[[n]]$se)
  dx = data.frame(x=x,y=y,ci.upper=se.upper,ci.lower=se.lower)
  plen = ggplot(dx, aes(x=x,y=y))+
    geom_ribbon(aes(ymin=ci.lower,ymax=ci.upper), alpha=0.3)+
   geom_line(size=1) +
    xlab(xlab)+
   ylab("Probability of borrowing")+
    coord_cartesian(ylim = c(0,1))
  return(plen)
}
```

Load data

The Dutch data is processed very similarly to the English data. The full process can be found in the processing folder, but here we just load the final prepared data frame:

```
load("../data/loanwords_Dutch.Rdat")
```

Part of speech

```
catx = data.frame(
 PoS = tapply(dutch$cat, dutch$cat, function(X){as.character(X[1])}),
 mean = tapply(dutch$bor15, dutch$cat, mean),
 n = tapply(dutch$bor15, dutch$cat, length)
catx = catx[order(catx$mean, decreasing = T),]
catx$PoS = factor(catx$PoS, levels = catx[order(catx$mean, decreasing = T),]$PoS)
posg = ggplot(catx, aes(x=mean, y=PoS)) +
  geom_point(size=2) +
 ylab("Part of speech") +
 xlab("Proportion of words borrowed")+
  scale_x_continuous(labels=percent_format()) +
  geom_text(aes(label=n), nudge_y=0.4)
pdf("../results/graphs/POS_Borrowing_Dutch.pdf",
   width = 6,
   height = 4)
posg
dev.off()
## pdf
catx$mean= catx$mean*100
write.csv(catx, "../results/Dutch_POS_BorrowingProportions.csv", row.names = F)
```

GAM model

Dutch data has 1028 datapoints.

The range of the length variable limits the number of knots that the gam model can fit:

summary(m0.dutch)

```
## Family: binomial
## Link function: logit
## Formula:
## bor15.cat ~ s(phonlengthscale, k = 3) + s(AoAscale) + s(subtlexzipfscale) +
      s(concscale) + s(cat, bs = "re") + s(cat, phonlengthscale,
      bs = "re") + s(cat, AoAscale, bs = "re") + s(cat, subtlexzipfscale,
##
##
      bs = "re") + s(cat, concscale, bs = "re")
##
## Parametric coefficients:
              Estimate Std. Error z value Pr(>|z|)
##
## (Intercept) -2.3389
                         0.3779 -6.189 6.04e-10 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Approximate significance of smooth terms:
                               edf Ref.df Chi.sq p-value
                         1.942e+00 1.996 18.147 0.000149 ***
## s(phonlengthscale)
                         1.310e+00 1.561 11.260 0.005448 **
## s(AoAscale)
## s(subtlexzipfscale) 3.630e+00 4.559 11.863 0.022505 *
## s(concscale)
                         1.663e+00 2.060 2.980 0.241790
## s(cat)
                         3.724e+00 10.000 39.054 1.98e-08 ***
## s(cat,phonlengthscale) 1.721e-01 10.000 0.194 0.293757
## s(cat, AoAscale) 9.855e-06 10.000 0.000 0.938779
## s(cat,subtlexzipfscale) 1.883e-05 10.000 0.000 0.669506
                         1.184e+00 10.000 3.069 0.098963 .
## s(cat,concscale)
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## R-sq.(adj) = 0.112 Deviance explained = 13.4\%
## fREML = 1435 Scale est. = 1 n = 1028
```

Interactions

Test whether an interaction between AoA and frequency is warranted:

```
##
## Model 1: bor15.cat ~ s(phonlengthscale, k = 3) + s(AoAscale) + s(subtlexzipfscale) +
       s(concscale) + s(cat, bs = "re") + s(cat, phonlengthscale,
##
       bs = "re") + s(cat, AoAscale, bs = "re") + s(cat, subtlexzipfscale,
       bs = "re") + s(cat, concscale, bs = "re")
##
## Model 2: bor15.cat ~ s(phonlengthscale, k = 3) + s(AoAscale) + s(subtlexzipfscale) +
       s(concscale) + s(cat, bs = "re") + s(cat, phonlengthscale,
##
##
       bs = "re") + s(cat, AoAscale, bs = "re") + s(cat, subtlexzipfscale,
##
       bs = "re") + s(cat, concscale, bs = "re") + te(AoAscale,
       subtlexzipfscale)
##
##
        #Df LogLik
                        Df Chisq Pr(>Chisq)
## 1 18.472 -441.31
## 2 20.348 -439.66 1.8758 3.2968
                                      0.1924
```

No significant improvement.

Test whether an interaction between AoA and length is warranted:

Likelihood ratio test

```
##
## Model 1: bor15.cat ~ s(phonlengthscale, k = 3) + s(AoAscale) + s(subtlexzipfscale) +
       s(concscale) + s(cat, bs = "re") + s(cat, phonlengthscale,
       bs = "re") + s(cat, AoAscale, bs = "re") + s(cat, subtlexzipfscale,
##
##
       bs = "re") + s(cat, concscale, bs = "re")
## Model 2: bor15.cat ~ s(phonlengthscale, k = 3) + s(AoAscale) + s(subtlexzipfscale) +
       s(concscale) + s(cat, bs = "re") + s(cat, phonlengthscale,
       bs = "re") + s(cat, AoAscale, bs = "re") + s(cat, subtlexzipfscale,
##
##
       bs = "re") + s(cat, concscale, bs = "re") + te(AoAscale,
##
       phonlengthscale)
        #Df LogLik
                        Df Chisq Pr(>Chisq)
## 1 18.472 -441.31
## 2 21.220 -439.07 2.7477 4.4682
                                      0.2151
```

There is no improvement in log likelihood.

Test whether an interaction between Frequency and length is warranted:

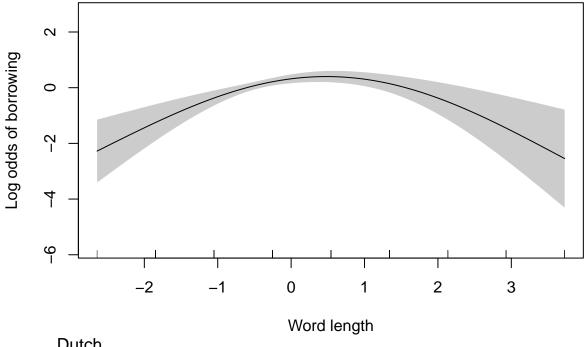
```
## Likelihood ratio test
## Model 1: bor15.cat ~ s(phonlengthscale, k = 3) + s(AoAscale) + s(subtlexzipfscale) +
       s(concscale) + s(cat, bs = "re") + s(cat, phonlengthscale,
##
       bs = "re") + s(cat, AoAscale, bs = "re") + s(cat, subtlexzipfscale,
##
##
       bs = "re") + s(cat, concscale, bs = "re")
## Model 2: bor15.cat ~ s(phonlengthscale, k = 3) + s(AoAscale) + s(subtlexzipfscale) +
       s(concscale) + s(cat, bs = "re") + s(cat, phonlengthscale,
##
##
       bs = "re") + s(cat, AoAscale, bs = "re") + s(cat, subtlexzipfscale,
       bs = "re") + s(cat, concscale, bs = "re") + te(subtlexzipfscale,
##
##
      phonlengthscale)
        #Df LogLik
##
                        Df Chisq Pr(>Chisq)
## 1 18.472 -441.31
## 2 22.040 -437.60 3.5681 7.4151
                                      0.1155
```

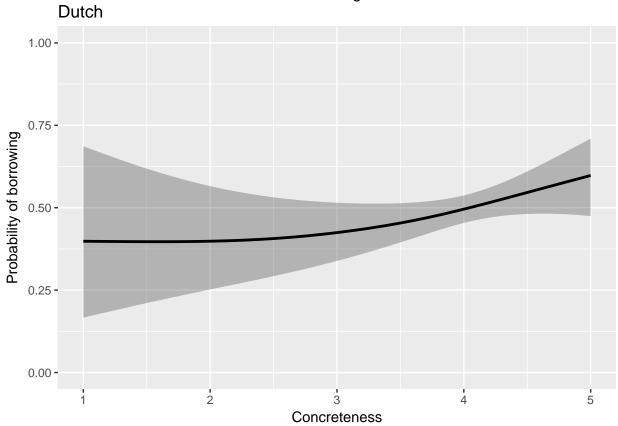
No significant improvement.

So no interactions are necessary.

Model estiamtes

Plot the model estimates, changing the dependent scale to probability and the independent variables to their original scales (code is hidden, but available in the Rmd file).

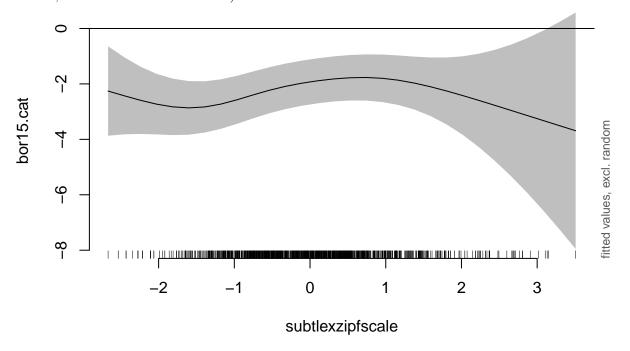


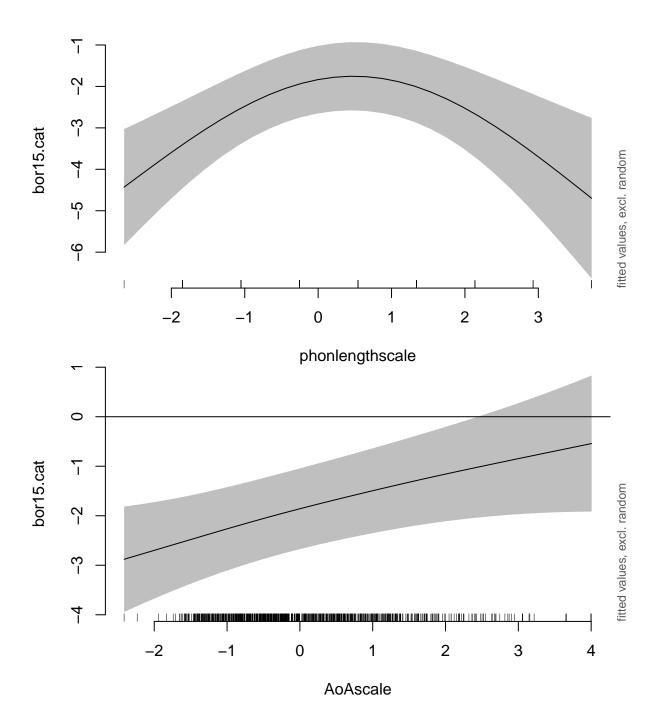


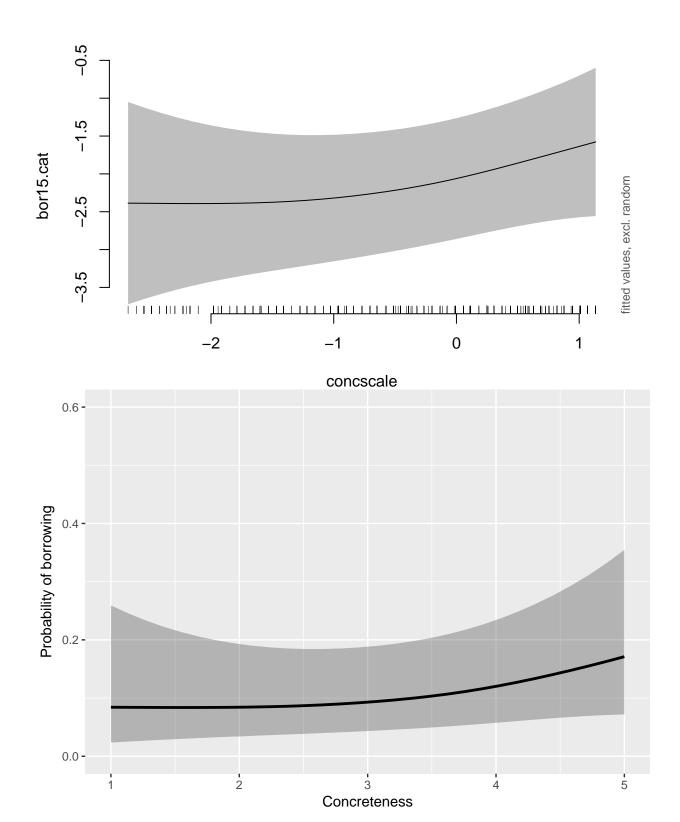


pdf ## 2

Plot the model estimates, removing the influence of the random effects using the library itsadug (code is hidden, but available in the Rmd file).









pdf ## 2

Random effects for Part of speech

```
mc = m0.dutch$coefficients
mc[grepl("s\\(cat\\)",names(mc))]
                    s(cat).2
                                  s(cat).3
                                               s(cat).4
##
       s(cat).1
                                                             s(cat).5
    1.119986827
                 0.421313755 -0.262839333 -0.037142605 -0.066124188
##
       s(cat).6
                    s(cat).7
                                  s(cat).8
                                               s(cat).9
                                                            s(cat).10
##
   -0.004943456 -0.103771803 -0.131386158 -0.241569599
                                                         0.101695214
##
      s(cat).11
##
## -0.795218653
raw = tapply(dutch$bor15,dutch$cat,mean)
model.est = logit2per(m0.dutch$coefficients[1] +
              mc[grepl("s\\(cat\\)",names(mc))])
plot(raw, model.est,
  xlab="Raw proportions",
  ylab="Model estimates",
  col="white",
     ylim=c(0,0.3),
     xlim=c(0,0.3))
abline(0,1,col='gray')
text(raw, model.est, names(raw))
```

