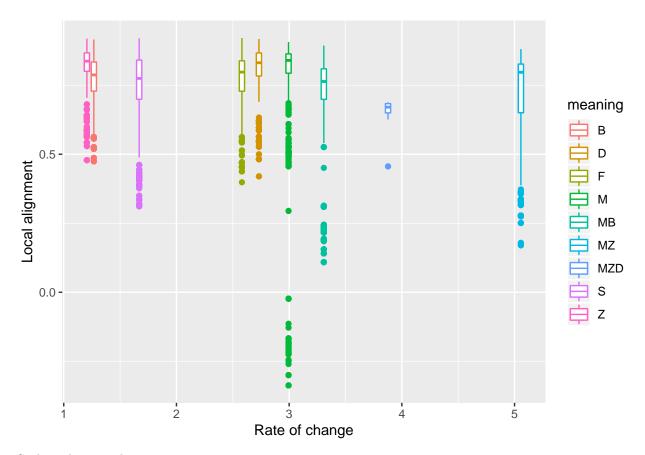
Rate of change for Kinship words

```
library(dplyr)
library(lme4)
library(ggplot2)
Read in data:
d = read.csv("../data/FAIR/nel-wiki-k100-alignments-merged-long.csv",
             encoding = "UTF-8",fileEncoding = "UTF-8",stringsAsFactors = F)
# Data on languages
1 = read.csv("../data/FAIR_langauges_glotto_xdid.csv",stringsAsFactors = F)
# Read in kinship rate of change data
r = read.csv("../data/RaczPassmoreSheardJordan_2019/supp-data-si.csv",
             encoding = "UTF-8",fileEncoding = "UTF-8",stringsAsFactors = F)
# Edit language isos
r$language.iso = l[match(r$language, l$Language), ]$iso2
r[r$language=="Norwegian",]$language.iso = "no"
r[r$language=="Ossetic",]$language.iso = "os"
# copy
rkin = r
Merge the data:
rkin2 = left join(d,rkin,by = c("Word Form 11"="word","11"="language.iso"))
names(rkin) = paste0(names(rkin),".12")
rkin2 = left_join(rkin2,rkin,by = c("Word_Form_12"="word.12","12"="language.iso.12"))
Exclude missing data:
rkin2 = rkin2[!is.na(rkin2$mean.roc),]
rkin2 = rkin2[!is.na(rkin2$mean.roc.12),]
Calculate difference in log frequency per million words:
rkin2$fpm.l1 = rkin2$word.count / (rkin2$corpora.size/1000000)
rkin2$fpm.12 = rkin2$word.count.12 / (rkin2$corpora.size.12/1000000)
rkin2$freq_diff = log10(abs(rkin2$fpm.l1 - rkin2$fpm.l2))
Filter variables:
rkin3 = rkin2[,c("l1","l2","meaning","Word_Form_l1","Word_Form_l2",
                 "local_alignment", "Concept_ID", "fpm.11", "fpm.12",
                 "mean.roc", "mean.roc.12", "lingpy.cognate", "expert.cognate",
                 "glottocode", "glottocode.12", "freq_diff")]
Plot data:
ggplot(rkin3, aes(y=local_alignment,x=mean.roc,color=meaning)) +
  geom_boxplot() +xlab("Rate of change") + ylab("Local alignment")
```



Scale and center data

Variable for language pair:

```
rkin3$langPair = apply(rkin3[,c("l1","l2")],1,function(X){
  paste(sort(X),collapse="-")
})
```

Predit local alignment by rate of change, with random intercepts for each cognate within each meaning, and random intercepts for l1 and l2. Note that we're predicting local alignment from rate of change (not the other way around). This is because rate of change is unique to a particular cognate within a particular meaning.

refitting model(s) with ML (instead of REML)

```
## Data: rkin3
## Models:
## m0: local_alignment.scaled ~ (1 | lingpy.cognate/meaning) + (1 |
           langPair)
## m1: local_alignment.scaled ~ (1 | lingpy.cognate/meaning) + (1 |
           langPair) + mean.roc.scaled
                BIC logLik deviance Chisq Chi Df Pr(>Chisq)
           AIC
## m0 5 20445 20482 -10217
                               20435
## m1 6 20436 20481 -10212
                               20424 10.222
                                                     0.001388 **
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
summary(m1)
## Linear mixed model fit by REML ['lmerMod']
## Formula: local_alignment.scaled ~ (1 | lingpy.cognate/meaning) + (1 |
##
       langPair) + mean.roc.scaled
##
      Data: rkin3
## Control: lmerControl(optimizer = "bobyqa")
## REML criterion at convergence: 20431.4
##
## Scaled residuals:
##
       Min
                  1Q Median
                                    3Q
                                            Max
## -13.4490 -0.2459 0.0084
                                0.2600
                                         4.0999
##
## Random effects:
## Groups
                           Name
                                       Variance Std.Dev.
## langPair
                           (Intercept) 0.7473
                                                0.8644
## lingpy.cognate
                           (Intercept) 0.2865
                                                0.5353
## meaning:lingpy.cognate (Intercept) 0.1086
                                                0.3295
## Residual
                                       0.2457
                                                0.4957
## Number of obs: 13061, groups:
## langPair, 275; lingpy.cognate, 98; meaning:lingpy.cognate, 98
##
## Fixed effects:
                   Estimate Std. Error t value
##
                   -0.48572
                               0.08586 -5.657
## (Intercept)
## mean.roc.scaled -0.19683
                               0.06037 -3.260
## Correlation of Fixed Effects:
               (Intr)
## men.rc.scld -0.102
Same as above, but also control for frequency difference (part of speech is the same).
rkin3Freq = rkin3[!is.na(rkin3$freq_diff),]
rkin3Freq$freq_diff.scaled = scale(rkin3Freq$freq_diff)
m0.freq = lmer(local_alignment.scaled ~ freq_diff.scaled +
            (1|lingpy.cognate/meaning) +
            (1|langPair),
          data=rkin3Freq,
          control = lmerControl(optimizer = "bobyqa"))
m1.freq = update(m0.freq, ~.+mean.roc.scaled)
anova(m0.freq,m1.freq)
```

```
## refitting model(s) with ML (instead of REML)
## Data: rkin3Freq
## Models:
## m0.freq: local_alignment.scaled ~ freq_diff.scaled + (1 | lingpy.cognate/meaning) +
                (1 | langPair)
## m0.freq:
## m1.freq: local_alignment.scaled ~ freq_diff.scaled + (1 | lingpy.cognate/meaning) +
## m1.freq:
                (1 | langPair) + mean.roc.scaled
##
           Df
                  AIC
                          BIC logLik deviance Chisq Chi Df Pr(>Chisq)
## m0.freq 6 -3643.2 -3599.7 1827.6 -3655.2
## m1.freq 7 -3644.6 -3593.8 1829.3 -3658.6 3.3989
                                                                0.06524 .
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
summary(m1.freq)
## Linear mixed model fit by REML ['lmerMod']
## local_alignment.scaled ~ freq_diff.scaled + (1 | lingpy.cognate/meaning) +
       (1 | langPair) + mean.roc.scaled
##
      Data: rkin3Freq
## Control: lmerControl(optimizer = "bobyqa")
##
## REML criterion at convergence: -3641.9
##
## Scaled residuals:
##
       Min
                1Q Median
                                30
                                       Max
## -5.1935 -0.5043 0.0438 0.5373 6.4121
##
## Random effects:
## Groups
                           Name
                                       Variance Std.Dev.
                           (Intercept) 0.17776 0.4216
## langPair
## lingpy.cognate
                           (Intercept) 0.12922 0.3595
## meaning:lingpy.cognate (Intercept) 0.44379 0.6662
## Residual
                                       0.03569 0.1889
## Number of obs: 10481, groups:
## langPair, 171; lingpy.cognate, 86; meaning:lingpy.cognate, 86
##
## Fixed effects:
##
                     Estimate Std. Error t value
                    -0.006664 0.089906 -0.074
## (Intercept)
## freq_diff.scaled -0.004499
                                0.002196 -2.048
## mean.roc.scaled -0.141280
                               0.076744 -1.841
##
## Correlation of Fixed Effects:
               (Intr) frq d.
## frq_dff.scl 0.003
## men.rc.scld -0.118 0.006
Plot data with model estimate (solid line), and estimate when taking frequency into account (dashed line).
The rate of change is calculated for each meaning, hence the clustering. Note the plot is in scaled space.
```

