Attempt to calculate LENA accuracy for ling/nonling

AC

Read data in

NOTE: This section cannot be reproduced because it requires access to data that haven't been shared.

```
read.csv("result_final_lisa.csv")->metadata
metadata[1,]
csvs=dir("../Raw_Data/its_to_csv/",pattern="csv")
#count_present=NULL
all_segments=NULL
# extract info from csv's
for(thisits in levels(metadata$filename)){
  csv=grep(thisits,csvs)
  #count_present=c(count_present, length(csv))
  if(length(csv)==1){
    #read in full csv
   read.csv(paste0(".../Raw_Data/its_to_csv/",csvs[csv]))->thisdat
    #select down to the lines used in zooniverse
    sel segments start=metadata[metadata$filename==thisits, "Starttime"]
    sel_segments_dat=thisdat[thisdat$startTime %in% sel_segments_start,]
    #the following is NOT right: we take the data in even if there is a mismatch in start/end time
    if(dim(sel_segments_dat)[1]!=length(sel_segments_start)) print(paste(thisits, csvs[csv], "have line
    if(dim(sel_segments_dat)[1]>0){
     #derive LENA classification at the segment level
   sel_segments_dat$LENA_type=NA
      #if ling info contained, then ling
   sel_segments_dat$LENA_type[is.na(sel_segments_dat$LENA_type) & sel_segments_dat$childUttLen>0 & !is
      #elsif there is crying
   sel_segments_dat$LENA_type[is.na(sel_segments_dat$LENA_type) & !is.na(sel_segments_dat$endCry1)]<-
     #elsif there is fixed then laughing
   sel_segments_dat$LENA_type[is.na(sel_segments_dat$LENA_type) & !is.na(sel_segments_dat$endVfx1)]<-
    #else junk
   sel_segments_dat$LENA_type[is.na(sel_segments_dat$LENA_type)]<-"Junk"
    sel_segments_dat=sel_segments_dat[,c("itsId","startTime", "endTime","LENA_type")]
    #combine with lab annotation
    sel_segments_dat=merge(sel_segments_dat,metadata[metadata$filename==thisits,],by.x="startTime",by.y
    #add to full spreadsheet
```

```
all_segments=rbind(all_segments,sel_segments_dat)
}

else print(paste(thisits,"not found"))
}
write.csv(all_segments,"LENA_type_lab.csv",row.names = F)
```

Correspondence between LENA & lab annotation at the level of segments

Here we look at to what extent LENA and lab annotations match at the level of individual segments. Each data point is one segment (one "vocalization").

```
read.csv("LENA_type_lab.csv")->all_segments
## NOOOOOOTE !!!! REMOVING Num_Agreement < 2
all_segments=all_segments[all_segments$Num_Agreement>=2,]
# create lab column with easier to read correspondance
all_segments\$lab<-as.character(all_segments\$Major_Choice)
all_segments$lab[all_segments$lab=="Non-canonical syllables"] <- "Non-Canonical"
all_segments$lab[all_segments$lab=="Canonical syllables"]<-"Canonical"
all_segments$lab[all_segments$lab=="Words"] <- "Canonical" #!!! why didn't I need this before??
all_segments$lab[all_segments$lab %in% c("Don't mark","None")]<-"Junk"
#collapse acros can and non-can
all_segments$lab[all_segments$lab %in% c("Non-Canonical", "Canonical")]<-"Can/non-can"
table(all_segments$lab)
##
## Can/non-can
                    Crying
                                  Junk
                                          Laughing
##
          8502
                       598
                                  2361
                                                188
table(all_segments$LENA_type)
## Can/non-can
                    Crying
                                  Junk
                                          Laughing
          8108
                      3028
                                                483
all_segments$lab=factor(all_segments$lab,levels=c("Can/non-can","Laughing","Crying","Junk"))
all_segments$LENA_type=factor(all_segments$LENA_type,levels=c("Can/non-can","Laughing","Crying","Junk")
mycf=confusionMatrix(all_segments$lab, all_segments$LENA_type, dnn = c("Lab","LENA"))
conf_tab=mycf$table
# this package uses sensitivity & specificity
#Sensitivity=recall
#Specificity=precision
mycf
## Confusion Matrix and Statistics
##
```

```
##
                 LENA
## Lab
                  Can/non-can Laughing Crying Junk
##
     Can/non-can
                         6213
                                    270
                                          2012
                           82
                                     22
                                            84
                                                  0
##
     Laughing
##
     Crying
                          126
                                      4
                                           467
                                                  1
     Junk
                         1687
                                           465
                                                 22
##
                                    187
##
## Overall Statistics
##
##
                   Accuracy : 0.5772
                     95% CI : (0.5682, 0.5862)
##
       No Information Rate: 0.696
##
       P-Value [Acc > NIR] : 1
##
##
##
                      Kappa: 0.1145
##
    Mcnemar's Test P-Value : <2e-16
##
##
## Statistics by Class:
##
##
                         Class: Can/non-can Class: Laughing Class: Crying
## Sensitivity
                                      0.7663
                                                    0.045549
                                                                     0.15423
## Specificity
                                      0.3536
                                                    0.985133
                                                                     0.98480
## Pos Pred Value
                                      0.7308
                                                    0.117021
                                                                     0.78094
## Neg Pred Value
                                      0.3978
                                                    0.959777
                                                                     0.76826
## Prevalence
                                      0.6960
                                                    0.041463
                                                                     0.25994
## Detection Rate
                                      0.5334
                                                    0.001889
                                                                     0.04009
## Detection Prevalence
                                                                     0.05133
                                      0.7298
                                                    0.016139
## Balanced Accuracy
                                      0.5599
                                                    0.515341
                                                                     0.56952
##
                         Class: Junk
## Sensitivity
                            0.733333
## Specificity
                            0.798692
## Pos Pred Value
                            0.009318
## Neg Pred Value
                            0.999139
## Prevalence
                            0.002575
## Detection Rate
                            0.001889
## Detection Prevalence
                            0.202678
## Balanced Accuracy
                            0.766013
```

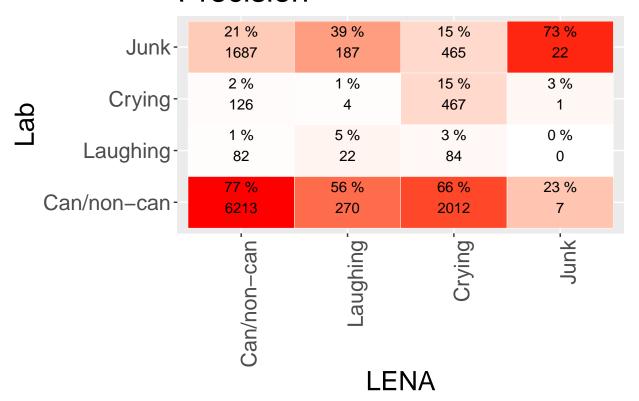
Precision

Precision means: If a segment was called X by LENA, what proportion of the time was it called X by lab coders?

```
colsums=colSums(conf_tab)
my_conf_tab=conf_tab
for(i in 1:dim(my_conf_tab)[2]) my_conf_tab[,i]=my_conf_tab[,i]/colsums[i]
colSums(my_conf_tab)

## Can/non-can Laughing Crying Junk
## 1 1 1 1
prop_cat=data.frame(my_conf_tab*100) #generates precision because columns
prop_cat$id=paste(prop_cat$Lab,prop_cat$LENA)
```

Precision



Recall

Recall means: If a segment was called X by lab coders, what proportion of the time was it called X by LENA coders?

```
prop_cat=data.frame(conf_tab/rowSums(conf_tab)*100) #generates recall because rows
prop_cat$id=paste(prop_cat$Lab,prop_cat$LENA)
colnames(prop_cat)[3]<-"rec"
data.frame(conf_tab)->stall
stall$id=paste(stall$Lab,stall$LENA)
stall=merge(stall,prop_cat[c("id","rec")])
ggplot(data = stall, mapping = aes(y = Lab, x=LENA)) +
```

Recall

