

Analyses for JSLHR version

2020-10-17

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History:

- 2020-08-05 final first version
- 2020-10-10 (rehaul),
- latest minor edits 2020-10-17

TODO:

- pipeline is still not transparent
- there are duplicate files across raw and derived data
- there are a bunch of files with similar names
- README is old
- a note said read demo data created by AC from info in paper - should be replaced with real demo data

Read data in

```
# read datasets

demo_data=read.csv("../Derived_Data/demo-data.tsv",sep="\t")
data_ang <- read.csv("../Derived_Data/classifications_PU_zoon_final17.csv",header=T,sep=",")
data_td <- read.csv("../Derived_Data/classifications_PU_zoon_final.csv")
data_all<-rbind(data_ang, data_td)

#add filenames to demo data
demo_data_fn <- demo_data %>%
  left_join(select(data_all, filename, ChildID), by = c("ChildID"))
```

```
## Warning: Column `ChildID` joining factors with different levels, coercing to
## character vector
demo_data_fn<-unique(demo_data_fn)

#remove the word mixed that takes up space and is unnecessary
data_all$Zoon_classif=factor(gsub("Mixed_", "", as.character(data_all$Zoon_classif), fixed=T))
#relevel the factor so that it's easier to read
data_all$Zoon_classif=factor(data_all$Zoon_classif, levels=c("Canonical", "Non-Canonical",
# create lab column with easier to read correspondance
data_all$lab<-as.character(data_all$Major_Choice)
data_all$lab[data_all$lab=="Non-canonical syllables"]<-"Non-Canonical"
data_all$lab[data_all$lab=="Canonical syllables"]<-"Canonical"
data_all$lab[data_all$lab %in% c("Don't mark", "None")]<-"Junk"
data_all$lab=factor(data_all$lab, levels=levels(data_all$Zoon_classif))
#apply same factor levels as zooniverse so that we can do symmetrical confusion matrices
```

Correspondence between lab & zooniverse annotation at the level of segments

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

```
table(data_all$lab)
```

```
##
##              Canonical              Non-Canonical
##              1779              6423
##              Crying              Laughing
##              588              186
##              Junk              Canonical_Crying
##              2595              0
##              Canonical_Laughing              Crying_Canonical
##              0              0
##              Crying_Canonical_Laughing              Crying_Laughing
##              0              0
##              Crying_Laughing_Non-Canonical              Crying_Non-Canonical
##              0              0
##              Crying_Non-Canonical_Laughing              Laughing_Crying
##              0              0
##              Laughing_Non-Canonical Laughing_Non-Canonical_Crying
##              0              0
##              Non-Canonical_Crying              Non-Canonical_Laughing
##              0              0
```

```
table(data_all$Zoon_classif)
```

```
##
##              Canonical              Non-Canonical
##              1665              5525
##              Crying              Laughing
##              920              442
##              Junk              Canonical_Crying
##              1456              3
##              Canonical_Laughing              Crying_Canonical
##              15              30
```

```
##      Crying_Canonical_Laughing      Crying_Laughing
##              2              108
## Crying_Laughing_Non-Canonical      Crying_Non-Canonical
##              75              655
## Crying_Non-Canonical_Laughing      Laughing_Crying
##              7              8
##      Laughing_Non-Canonical Laughing_Non-Canonical_Crying
##              124              9
##      Non-Canonical_Crying      Non-Canonical_Laughing
##              338              211
```

```
mycf=confusionMatrix(data_all$lab, data_all$Zoon_classif, dnn = c("Lab","Zooniverse"))
conf_tab=mycf$table
# this package uses sensitivity & specificity
#Sensitivity=recall
#Specificity=precision
mycf
```

```
## Confusion Matrix and Statistics
```

```
##
##              Zooniverse
## Lab      Canonical Non-Canonical Crying Laughing Junk
## Canonical      1014      524      31      29      79
## Non-Canonical      255      4176      536      154      360
## Crying              6       32      239       8       6
## Laughing           2       12       8      115       6
## Junk             372      776      106      136      1005
## Canonical_Crying      0       0       0       0       0
## Canonical_Laughing     0       0       0       0       0
## Crying_Canonical      0       0       0       0       0
## Crying_Canonical_Laughing 0       0       0       0       0
## Crying_Laughing       0       0       0       0       0
## Crying_Laughing_Non-Canonical 0       0       0       0       0
## Crying_Non-Canonical  0       0       0       0       0
## Crying_Non-Canonical_Laughing 0       0       0       0       0
## Laughing_Crying       0       0       0       0       0
## Laughing_Non-Canonical 0       0       0       0       0
## Laughing_Non-Canonical_Crying 0       0       0       0       0
## Non-Canonical_Crying  0       0       0       0       0
## Non-Canonical_Laughing 0       0       0       0       0
```

```
##              Zooniverse
## Lab      Canonical_Crying Canonical_Laughing
## Canonical      0          2
## Non-Canonical      3          5
## Crying          0          0
## Laughing        0          0
## Junk           0          8
## Canonical_Crying 0          0
## Canonical_Laughing 0          0
## Crying_Canonical 0          0
## Crying_Canonical_Laughing 0          0
## Crying_Laughing   0          0
## Crying_Laughing_Non-Canonical 0          0
## Crying_Non-Canonical 0          0
## Crying_Non-Canonical_Laughing 0          0
```

##	Laughing_Crying	0	0
##	Laughing_Non-Canonical	0	0
##	Laughing_Non-Canonical_Crying	0	0
##	Non-Canonical_Crying	0	0
##	Non-Canonical_Laughing	0	0
##		Zooniverse	
##	Lab	Crying_Canonical	Crying_Canonical_Laughing
##	Canonical	21	2
##	Non-Canonical	5	0
##	Crying	1	0
##	Laughing	0	0
##	Junk	2	0
##	Canonical_Crying	0	0
##	Canonical_Laughing	0	0
##	Crying_Canonical	0	0
##	Crying_Canonical_Laughing	0	0
##	Crying_Laughing	0	0
##	Crying_Laughing_Non-Canonical	0	0
##	Crying_Non-Canonical	0	0
##	Crying_Non-Canonical_Laughing	0	0
##	Laughing_Crying	0	0
##	Laughing_Non-Canonical	0	0
##	Laughing_Non-Canonical_Crying	0	0
##	Non-Canonical_Crying	0	0
##	Non-Canonical_Laughing	0	0
##		Zooniverse	
##	Lab	Crying_Laughing	Crying_Laughing_Non-Canonical
##	Canonical	2	2
##	Non-Canonical	32	16
##	Crying	48	50
##	Laughing	11	1
##	Junk	15	6
##	Canonical_Crying	0	0
##	Canonical_Laughing	0	0
##	Crying_Canonical	0	0
##	Crying_Canonical_Laughing	0	0
##	Crying_Laughing	0	0
##	Crying_Laughing_Non-Canonical	0	0
##	Crying_Non-Canonical	0	0
##	Crying_Non-Canonical_Laughing	0	0
##	Laughing_Crying	0	0
##	Laughing_Non-Canonical	0	0
##	Laughing_Non-Canonical_Crying	0	0
##	Non-Canonical_Crying	0	0
##	Non-Canonical_Laughing	0	0
##		Zooniverse	
##	Lab	Crying_Non-Canonical	
##	Canonical	34	
##	Non-Canonical	420	
##	Crying	165	
##	Laughing	1	
##	Junk	35	
##	Canonical_Crying	0	
##	Canonical_Laughing	0	

##	Crying_Canonical	0	
##	Crying_Canonical_Laughing	0	
##	Crying_Laughing	0	
##	Crying_Laughing_Non-Canonical	0	
##	Crying_Non-Canonical	0	
##	Crying_Non-Canonical_Laughing	0	
##	Laughing_Crying	0	
##	Laughing_Non-Canonical	0	
##	Laughing_Non-Canonical_Crying	0	
##	Non-Canonical_Crying	0	
##	Non-Canonical_Laughing	0	
##			Zooniverse
##	Lab		Crying_Non-Canonical_Laughing Laughing_Crying
##	Canonical	1	1
##	Non-Canonical	4	5
##	Crying	1	1
##	Laughing	0	0
##	Junk	1	1
##	Canonical_Crying	0	0
##	Canonical_Laughing	0	0
##	Crying_Canonical	0	0
##	Crying_Canonical_Laughing	0	0
##	Crying_Laughing	0	0
##	Crying_Laughing_Non-Canonical	0	0
##	Crying_Non-Canonical	0	0
##	Crying_Non-Canonical_Laughing	0	0
##	Laughing_Crying	0	0
##	Laughing_Non-Canonical	0	0
##	Laughing_Non-Canonical_Crying	0	0
##	Non-Canonical_Crying	0	0
##	Non-Canonical_Laughing	0	0
##			Zooniverse
##	Lab		Laughing_Non-Canonical
##	Canonical	18	
##	Non-Canonical	62	
##	Crying	5	
##	Laughing	9	
##	Junk	30	
##	Canonical_Crying	0	
##	Canonical_Laughing	0	
##	Crying_Canonical	0	
##	Crying_Canonical_Laughing	0	
##	Crying_Laughing	0	
##	Crying_Laughing_Non-Canonical	0	
##	Crying_Non-Canonical	0	
##	Crying_Non-Canonical_Laughing	0	
##	Laughing_Crying	0	
##	Laughing_Non-Canonical	0	
##	Laughing_Non-Canonical_Crying	0	
##	Non-Canonical_Crying	0	
##	Non-Canonical_Laughing	0	
##			Zooniverse
##	Lab		Laughing_Non-Canonical_Crying
##	Canonical	0	

```

## Non-Canonical 4
## Crying 3
## Laughing 1
## Junk 1
## Canonical_Crying 0
## Canonical_Laughing 0
## Crying_Canonical 0
## Crying_Canonical_Laughing 0
## Crying_Laughing 0
## Crying_Laughing_Non-Canonical 0
## Crying_Non-Canonical 0
## Crying_Non-Canonical_Laughing 0
## Laughing_Crying 0
## Laughing_Non-Canonical 0
## Laughing_Non-Canonical_Crying 0
## Non-Canonical_Crying 0
## Non-Canonical_Laughing 0
##
## Zooniverse
## Lab Non-Canonical_Crying Non-Canonical_Laughing
## Canonical 8 11
## Non-Canonical 271 115
## Crying 20 3
## Laughing 0 20
## Junk 39 62
## Canonical_Crying 0 0
## Canonical_Laughing 0 0
## Crying_Canonical 0 0
## Crying_Canonical_Laughing 0 0
## Crying_Laughing 0 0
## Crying_Laughing_Non-Canonical 0 0
## Crying_Non-Canonical 0 0
## Crying_Non-Canonical_Laughing 0 0
## Laughing_Crying 0 0
## Laughing_Non-Canonical 0 0
## Laughing_Non-Canonical_Crying 0 0
## Non-Canonical_Crying 0 0
## Non-Canonical_Laughing 0 0
##
## Overall Statistics
##
## Accuracy : 0.566
## 95% CI : (0.5569, 0.575)
## No Information Rate : 0.4771
## P-Value [Acc > NIR] : < 2.2e-16
##
## Kappa : 0.3621
##
## McNemar's Test P-Value : NA
##
## Statistics by Class:
##
## Class: Canonical Class: Non-Canonical Class: Crying
## Sensitivity 0.61492 0.7565 0.25978
## Specificity 0.92290 0.6287 0.96723

```

## Pos Pred Value	0.56998	0.6502	0.40646
## Neg Pred Value	0.93515	0.7389	0.93800
## Prevalence	0.14251	0.4771	0.07951
## Detection Rate	0.08763	0.3609	0.02066
## Detection Prevalence	0.15375	0.5551	0.05082
## Balanced Accuracy	0.76891	0.6926	0.61351
##	Class: Laughing	Class: Junk	Class: Canonical_Crying
## Sensitivity	0.260181	0.69025	0.0000000
## Specificity	0.993620	0.84281	1.0000000
## Pos Pred Value	0.618280	0.38728	NaN
## Neg Pred Value	0.971278	0.94975	0.9997407
## Prevalence	0.038199	0.12583	0.0002593
## Detection Rate	0.009939	0.08686	0.0000000
## Detection Prevalence	0.016075	0.22427	0.0000000
## Balanced Accuracy	0.626901	0.76653	0.5000000
##	Class: Canonical_Laughing	Class: Crying_Canonical	
## Sensitivity	0.000000	0.000000	
## Specificity	1.000000	1.000000	
## Pos Pred Value	NaN	NaN	
## Neg Pred Value	0.998704	0.997494	
## Prevalence	0.001296	0.002506	
## Detection Rate	0.000000	0.000000	
## Detection Prevalence	0.000000	0.000000	
## Balanced Accuracy	0.500000	0.500000	
##	Class: Crying_Canonical_Laughing	Class: Crying_Laughing	
## Sensitivity	0.0000000	0.000000	
## Specificity	1.0000000	1.000000	
## Pos Pred Value	NaN	NaN	
## Neg Pred Value	0.9998272	0.990666	
## Prevalence	0.0001728	0.009334	
## Detection Rate	0.0000000	0.000000	
## Detection Prevalence	0.0000000	0.000000	
## Balanced Accuracy	0.5000000	0.500000	
##	Class: Crying_Laughing_Non-Canonical		
## Sensitivity	0.000000		
## Specificity	1.000000		
## Pos Pred Value	NaN		
## Neg Pred Value	0.993518		
## Prevalence	0.006482		
## Detection Rate	0.000000		
## Detection Prevalence	0.000000		
## Balanced Accuracy	0.500000		
##	Class: Crying_Non-Canonical		
## Sensitivity	0.00000		
## Specificity	1.00000		
## Pos Pred Value	NaN		
## Neg Pred Value	0.94339		
## Prevalence	0.05661		
## Detection Rate	0.00000		
## Detection Prevalence	0.00000		
## Balanced Accuracy	0.50000		
##	Class: Crying_Non-Canonical_Laughing		
## Sensitivity	0.000000		
## Specificity	1.000000		

```

## Pos Pred Value          NaN
## Neg Pred Value          0.999395
## Prevalence              0.000605
## Detection Rate          0.000000
## Detection Prevalence    0.000000
## Balanced Accuracy       0.500000
##
##      Class: Laughing_Crying Class: Laughing_Non-Canonical
## Sensitivity              0.0000000          0.00000
## Specificity              1.0000000          1.00000
## Pos Pred Value          NaN              NaN
## Neg Pred Value          0.9993086          0.98928
## Prevalence              0.0006914          0.01072
## Detection Rate          0.0000000          0.00000
## Detection Prevalence    0.0000000          0.00000
## Balanced Accuracy       0.5000000          0.50000
##
##      Class: Laughing_Non-Canonical_Crying
## Sensitivity              0.0000000
## Specificity              1.0000000
## Pos Pred Value          NaN
## Neg Pred Value          0.9992222
## Prevalence              0.0007778
## Detection Rate          0.0000000
## Detection Prevalence    0.0000000
## Balanced Accuracy       0.5000000
##
##      Class: Non-Canonical_Crying Class: Non-Canonical_Laughing
## Sensitivity              0.00000          0.00000
## Specificity              1.00000          1.00000
## Pos Pred Value          NaN              NaN
## Neg Pred Value          0.97079          0.98176
## Prevalence              0.02921          0.01824
## Detection Rate          0.00000          0.00000
## Detection Prevalence    0.00000          0.00000
## Balanced Accuracy       0.50000          0.50000

```

Precision

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

```

colsums=colSums(conf_tab)
my_conf_tab=conf_tab
for(i in 1:18) my_conf_tab[,i]=my_conf_tab[,i]/colsums[i]
colSums(my_conf_tab)

```

```

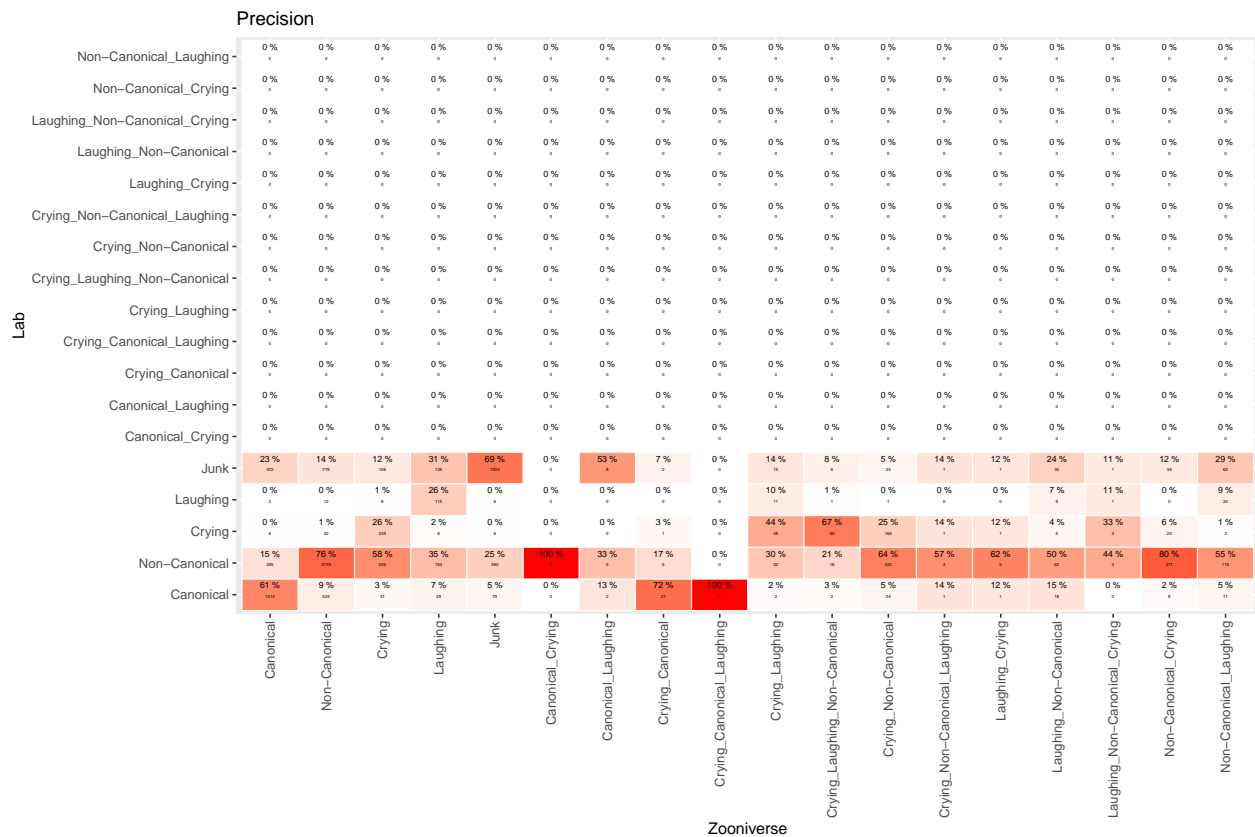
##      Canonical      Non-Canonical
##      1              1
##      Crying        Laughing
##      1              1
##      Junk          Canonical_Crying
##      1              1
##      Canonical_Laughing Crying_Canonical
##      1              1
##      Crying_Canonical_Laughing Crying_Laughing
##      1              1

```



```
## Crying_Laughing_Non-Canonical      Crying_Non-Canonical
##                                1                                1
## Crying_Non-Canonical_Laughing      Laughing_Crying
##                                1                                1
##      Laughing_Non-Canonical Laughing_Non-Canonical_Crying
##                                1                                1
##      Non-Canonical_Crying      Non-Canonical_Laughing
##                                1                                1
```

```
prop_cat=data.frame(my_conf_tab*100) #generates precision because columns
prop_cat$id=paste(prop_cat$Lab,prop_cat$Zooniverse)
colnames(prop_cat)[3]<-"pr"
data.frame(conf_tab)->stall
stall$id=paste(stall$Lab,stall$Zooniverse)
stall=merge(stall,prop_cat[c("id","pr")])
ggplot(data = stall, mapping = aes(y = Lab, x=Zooniverse)) +
  geom_tile(aes(fill= rescale(pr)), colour = "white") +
  geom_text(aes(label = paste(round(pr,"%")), vjust = -1,size=2) +
  geom_text(aes(label = Freq, vjust = 1,size=1) +
  scale_fill_gradient(low = "white", high = "red", name = "Percentage") +
  theme(legend.position = "none") +
  xlab("Zooniverse") + ylab("Lab") +
  ggtitle("Precision")+ theme(axis.text.x = element_text(angle = 90, vjust = 0.5, hjust=1))
```



Recall

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

```
rowsums=rowSums(conf_tab)
my_conf_tab=conf_tab
for(i in 1:18) my_conf_tab[,i]=my_conf_tab[,i]/rowsums[i]
rowSums(my_conf_tab)
```

```
##              Canonical              Non-Canonical
##              NaN                      NaN
##              Crying                    Laughing
##              NaN                      NaN
##              Junk                      Canonical_Crying
##              NaN                      NaN
##              Canonical_Laughing        Crying_Canonical
##              NaN                      NaN
##              Crying_Canonical_Laughing Crying_Laughing
##              NaN                      NaN
##              Crying_Laughing_Non-Canonical Crying_Non-Canonical
##              NaN                      NaN
##              Crying_Non-Canonical_Laughing Laughing_Crying
##              NaN                      NaN
##              Laughing_Non-Canonical Laughing_Non-Canonical_Crying
##              NaN                      NaN
##              Non-Canonical_Crying      Non-Canonical_Laughing
##              NaN                      NaN
```

```
prop_cat=data.frame(conf_tab/rowSums(conf_tab)*100) #generates recall because rows
prop_cat$id=paste(prop_cat$Lab,prop_cat$Zooniverse)
colnames(prop_cat)[3]<-"rec"
data.frame(conf_tab)->stall
stall$id=paste(stall$Lab,stall$Zooniverse)
stall=merge(stall,prop_cat[c("id","rec")])
ggplot(data = stall, mapping = aes(y = Lab, x=Zooniverse)) +
  geom_tile(aes(fill= rescale(rec)), colour = "white") +
  geom_text(aes(label = paste(round(rec),"%")), vjust = -1,size=2) +
  geom_text(aes(label = Freq), vjust = 1,size=1) +
  scale_fill_gradient(low = "white", high = "red", name = "Percentage") +
  theme(legend.position = "none") +
  xlab("Zooniverse") + ylab("Lab") +
  ggtitle("Recall")+ theme(axis.text.x = element_text(angle = 90, vjust = 0.5, hjust=1))
```

Lab		Recall															
		Canonical	Non-Canonical	Crying	Laughing	Junk	Canonical_Crying	Canonical_Laughing	Crying_Canonical	Crying_Canonical_Laughing	Crying_Laughing	Crying_Laughing_Non-Canonical	Crying_Non-Canonical	Laughing_Crying	Laughing_Non-Canonical	Laughing_Non-Canonical_Crying	Non-Canonical_Crying
Lab	Non-Canonical_Laughing -	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %
	Non-Canonical_Crying -	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %
	Laughing_Non-Canonical_Crying -	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %
	Laughing_Non-Canonical -	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %
	Laughing_Crying -	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %
	Crying_Non-Canonical_Laughing -	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %
	Crying_Non-Canonical -	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %
	Crying_Laughing_Non-Canonical -	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %
	Crying_Laughing -	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %
	Crying_Canonical_Laughing -	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %
	Crying_Canonical -	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %
	Canonical_Laughing -	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %
	Canonical_Crying -	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %	NaN %
	Junk -	14 % 22	30 % 176	4 % 104	5 % 124	39 % 1033	0 % 0	0 % 0	0 % 0	0 % 0	1 % 14	0 % 0	1 % 14	0 % 1	0 % 1	1 % 20	2 % 42
	Laughing -	1 % 2	6 % 12	4 % 8	62 % 119	3 % 4	0 % 0	0 % 0	0 % 0	0 % 0	6 % 11	1 % 2	0 % 0	0 % 0	5 % 9	1 % 2	11 % 20
	Crying -	1 % 4	5 % 12	41 % 103	1 % 4	1 % 4	0 % 0	0 % 0	0 % 0	0 % 0	8 % 14	9 % 16	28 % 60	0 % 1	0 % 1	1 % 2	3 % 6
	Non-Canonical -	4 % 88	65 % 1213	8 % 164	2 % 34	6 % 100	0 % 0	0 % 0	0 % 0	0 % 0	0 % 0	0 % 0	7 % 14	0 % 0	0 % 0	1 % 2	4 % 77
	Canonical -	87 % 104	29 % 104	2 % 21	2 % 21	4 % 10	0 % 0	0 % 0	1 % 2	0 % 0	0 % 0	0 % 0	2 % 4	0 % 0	0 % 0	1 % 2	1 % 11
		Canonical	Non-Canonical	Crying	Laughing	Junk	Canonical_Crying	Canonical_Laughing	Crying_Canonical	Crying_Canonical_Laughing	Crying_Laughing	Crying_Laughing_Non-Canonical	Crying_Non-Canonical	Laughing_Crying	Laughing_Non-Canonical	Laughing_Non-Canonical_Crying	Non-Canonical_Crying
		Zooniverse															

repeat collapsing

```

#given results above, we map the mixed
data_all$Zoon_classif[data_all$Zoon_classif=="Laughing_Canonical"]<-"Canonical"
data_all$Zoon_classif[data_all$Zoon_classif=="Laughing_Non-Canonical"]<-"Non-Canonical"
data_all$Zoon_classif[data_all$Zoon_classif=="Laughing_Non-Canonical_Crying"]<-"Non-Canonical"
data_all$Zoon_classif[data_all$Zoon_classif=="Laughing_Crying"]<-"Crying"
data_all$Zoon_classif[data_all$Zoon_classif=="Non-Canonical_Crying"]<-"Non-Canonical"
data_all$Zoon_classif[data_all$Zoon_classif=="Non-Canonical_Laughing_Crying"]<-"Non-Canonical"
data_all$Zoon_classif[data_all$Zoon_classif=="Crying_Canonical"]<-"Canonical"
# +
data_all$Zoon_classif[data_all$Zoon_classif=="Canonical_Crying"]<-"Canonical"
data_all$Zoon_classif[data_all$Zoon_classif=="Canonical_Laughing"]<-"Canonical"
data_all$Zoon_classif[data_all$Zoon_classif=="Laughing_Canonical_Crying"]<-"Non-Canonical"

data_all$Zoon_classif[data_all$Zoon_classif=="Crying_Laughing"]<-"Crying"
data_all$Zoon_classif[data_all$Zoon_classif=="Crying_Canonical_Laughing"]<-"Canonical"
data_all$Zoon_classif[data_all$Zoon_classif=="Crying_Laughing_Non-Canonical"]<-"Non-Canonical"
data_all$Zoon_classif[data_all$Zoon_classif=="Crying_Non-Canonical"]<-"Non-Canonical"
data_all$Zoon_classif[data_all$Zoon_classif=="Crying_Non-Canonical_Laughing"]<-"Non-Canonical"
data_all$Zoon_classif[data_all$Zoon_classif=="Non-Canonical_Laughing"]<-"Non-Canonical"

#and reset the factors for cleanliness
data_all$Zoon_classif=factor(data_all$Zoon_classif)
data_all$lab=factor(data_all$lab)
sample_data<-cbind(data_all$lab,data_all$Zoon_classif)

```

```

gac(data = sample_data, kat = 5, weight = c("unweighted"),
    conf.level = 0.95)

## Call:
## gac(data = sample_data, kat = 5, weight = c("unweighted"), conf.level = 0.95)
##
##      Estimate      StdErr   LowerCB UpperCB
## Const 0.5866689 0.0053816 0.5761201 0.5972
##
## Confidence level = 95%
## Sample size = 11571

mycf=confusionMatrix(data_all$lab, data_all$Zoon_classif, dnn = c("Lab","Zooniverse"))
conf_tab=mycf$table
# this package uses sensitivity & specificity
#Sensitivity=recall
#Specificity=precision
mycf

## Confusion Matrix and Statistics
##
##              Zooniverse
## Lab      Canonical Non-Canonical Crying Laughing Junk
## Canonical      1039           598         34         29         79
## Non-Canonical    268          5068        573        154        360
## Crying              7          279        288           8           6
## Laughing           2           44         19        115           6
## Junk             382          950        122        136       1005
##
## Overall Statistics
##
##              Accuracy : 0.6495
##              95% CI : (0.6407, 0.6582)
##      No Information Rate : 0.5997
##      P-Value [Acc > NIR] : < 2.2e-16
##
##              Kappa : 0.4265
##
## Mcnemar's Test P-Value : < 2.2e-16
##
## Statistics by Class:
##
##              Class: Canonical Class: Non-Canonical Class: Crying
## Sensitivity              0.61190              0.7304              0.27799
## Specificity              0.92505              0.7075              0.97152
## Pos Pred Value           0.58404              0.7890              0.48980
## Neg Pred Value           0.93270              0.6366              0.93189
## Prevalence               0.14675              0.5997              0.08953
## Detection Rate           0.08979              0.4380              0.02489
## Detection Prevalence     0.15375              0.5551              0.05082
## Balanced Accuracy        0.76847              0.7189              0.62476
##
##              Class: Laughing Class: Junk
## Sensitivity              0.260181              0.69025
## Specificity              0.993620              0.84281

```

```
## Pos Pred Value      0.618280      0.38728
## Neg Pred Value      0.971278      0.94975
## Prevalence          0.038199      0.12583
## Detection Rate      0.009939      0.08686
## Detection Prevalence 0.016075      0.22427
## Balanced Accuracy    0.626901      0.76653
```

```
pdf("../Results/precision_final.pdf",height=10,width=10)
colsums=colSums(conf_tab)
my_conf_tab=conf_tab
for(i in 1:5) my_conf_tab[,i]=my_conf_tab[,i]/colsums[i]
colSums(my_conf_tab)
```

```
##      Canonical Non-Canonical      Crying      Laughing      Junk
##              1              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$Zooniverse)
colnames(prop_cat)[3]<-"pr"
data.frame(conf_tab)->stall
stall$id=paste(stall$Lab,stall$Zooniverse)
stall=merge(stall,prop_cat[c("id","pr")])
ggplot(data = stall, mapping = aes(y = Lab, x=Zooniverse)) +
  geom_tile(aes(fill= rescale(pr)), colour = "white") +
  geom_text(aes(label = paste(round(pr,"%")), vjust = -1,size=8) +
  geom_text(aes(label = Freq), vjust = 1,size=8) +
  scale_fill_gradient(low = "white", high = "red", name = "Proportion") +
  theme(legend.position = "none") +
  xlab("Zooniverse") + ylab("Lab") +
  ggtitle("Precision")+theme(text = element_text(size=20),
    axis.text.x = element_text(angle=90, hjust=1))
dev.off()
```

```
## pdf
## 2
```

```
pdf("../Results/recall_final.pdf",height=10,width=10)
prop_cat=data.frame(conf_tab/rowSums(conf_tab)*100) #generates recall because rows
prop_cat$id=paste(prop_cat$Lab,prop_cat$Zooniverse)
colnames(prop_cat)[3]<-"rec"
data.frame(conf_tab)->stall
stall$id=paste(stall$Lab,stall$Zooniverse)
stall=merge(stall,prop_cat[c("id","rec")])
ggplot(data = stall, mapping = aes(y = Lab, x=Zooniverse)) +
  geom_tile(aes(fill= rescale(rec)), colour = "white") +
  geom_text(aes(label = paste(round(rec,"%")), vjust = -1,size=8) +
  geom_text(aes(label = Freq), vjust = 1,size=8) +
  scale_fill_gradient(low = "white", high = "red", name = "Proportion") +
  theme(legend.position = "none") +
  xlab("Zooniverse") + ylab("Lab") +
  ggtitle("Recall")+theme(text = element_text(size=20),
    axis.text.x = element_text(angle=90, hjust=1))
dev.off()
```

```
## pdf
## 2
```

```

data_as_td<-left_join(data_all,demo_data,on="ChildID")

## Joining, by = c("ChildID", "Age")

## Warning: Column `ChildID` joining factors with different levels, coercing to
## character vector

# CM with just AS kids
data_AS<-subset(data_as_td, Diagnosis=="AngelmanSyndrome")
mycf=confusionMatrix(data_AS$lab, data_AS$Zoon_classif, dnn = c("Lab","Zooniverse"))
conf_tab=mycf$table
mycf

```

```
## Confusion Matrix and Statistics
```

```
##
##              Zooniverse
## Lab      Canonical Non-Canonical Crying Laughing Junk
## Canonical           90          165           2          13          15
## Non-Canonical       100          2984          116          92         115
## Crying                1           39           15           2           1
## Laughing              0           16            4          57           2
## Junk                 239          618           34          92         462
```

```
## Overall Statistics
```

```
##
##              Accuracy : 0.6841
##              95% CI : (0.6714, 0.6966)
##      No Information Rate : 0.7247
##      P-Value [Acc > NIR] : 1
```

```
##
##              Kappa : 0.3624
```

```
##
## McNemar's Test P-Value : <2e-16
```

```
##
## Statistics by Class:
```

```
##
##              Class: Canonical Class: Non-Canonical Class: Crying
## Sensitivity           0.20930           0.7807           0.087719
## Specificity           0.95974           0.7087           0.991574
## Pos Pred Value        0.31579           0.8758           0.258621
## Neg Pred Value        0.93185           0.5512           0.970092
## Prevalence            0.08153           0.7247           0.032423
## Detection Rate        0.01706           0.5658           0.002844
## Detection Prevalence  0.05404           0.6460           0.010997
## Balanced Accuracy      0.58452           0.7447           0.539646
```

```
##
##              Class: Laughing Class: Junk
## Sensitivity           0.22266           0.7765
## Specificity           0.99562           0.7899
## Pos Pred Value        0.72152           0.3197
## Neg Pred Value        0.96169           0.9653
## Prevalence            0.04854           0.1128
## Detection Rate        0.01081           0.0876
## Detection Prevalence  0.01498           0.2740
## Balanced Accuracy      0.60914           0.7832
```

```
pdf("../Results/precision_AS.pdf",height=10,width=10)
colsums=colSums(conf_tab)
my_conf_tab=conf_tab
for(i in 1:5) my_conf_tab[,i]=my_conf_tab[,i]/colsums[i]
colSums(my_conf_tab)
```

```
##      Canonical Non-Canonical      Crying      Laughing      Junk
##      1          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$Zooniverse)
colnames(prop_cat)[3]<-"pr"
data.frame(conf_tab)->stall
stall$id=paste(stall$Lab,stall$Zooniverse)
stall=merge(stall,prop_cat[c("id","pr")])
ggplot(data = stall, mapping = aes(y = Lab, x=Zooniverse)) +
  geom_tile(aes(fill= rescale(pr)), colour = "white") +
  geom_text(aes(label = paste(round(pr,"%")), vjust = -1,size=8) +
  geom_text(aes(label = Freq), vjust = 1,size=8) +
  scale_fill_gradient(low = "white", high = "red", name = "Proportion") +
  theme(legend.position = "none") +
  xlab("Zooniverse") + ylab("Lab") +
  ggtitle("Precision")+theme(text = element_text(size=20),
    axis.text.x = element_text(angle=90, hjust=1))
dev.off()
```

```
## pdf
## 2
```

```
pdf("../Results/recall_AS.pdf",height=10,width=10)
prop_cat=data.frame(conf_tab/rowSums(conf_tab)*100) #generates recall because rows
prop_cat$id=paste(prop_cat$Lab,prop_cat$Zooniverse)
colnames(prop_cat)[3]<-"rec"
data.frame(conf_tab)->stall
stall$id=paste(stall$Lab,stall$Zooniverse)
stall=merge(stall,prop_cat[c("id","rec")])
ggplot(data = stall, mapping = aes(y = Lab, x=Zooniverse)) +
  geom_tile(aes(fill= rescale(rec)), colour = "white") +
  geom_text(aes(label = paste(round(rec,"%")), vjust = -1,size=8) +
  geom_text(aes(label = Freq), vjust = 1,size=8) +
  scale_fill_gradient(low = "white", high = "red", name = "Proportion") +
  theme(legend.position = "none") +
  xlab("Zooniverse") + ylab("Lab") +
  ggtitle("Recall")+theme(text = element_text(size=20),
    axis.text.x = element_text(angle=90, hjust=1))
dev.off()
```

```
## pdf
## 2
```

```
# CM with just TD kids
data_TD<-subset(data_as_td, Diagnosis=="Low-RiskControl")
mycf=confusionMatrix(data_TD$lab, data_TD$Zoon_classif, dnn = c("Lab","Zooniverse"))
conf_tab=mycf$table
mycf
```

```
## Confusion Matrix and Statistics
##
##              Zooniverse
## Lab          Canonical Non-Canonical Crying Laughing Junk
## Canonical          949          433          32          16          64
## Non-Canonical      168          2084          457          62          245
## Crying              6          240          273           6           5
## Laughing            2           28           15          58           4
## Junk              143          332           88          44          543
##
## Overall Statistics
##
##              Accuracy : 0.6205
##              95% CI : (0.6083, 0.6325)
##      No Information Rate : 0.495
##      P-Value [Acc > NIR] : < 2.2e-16
##
##              Kappa : 0.4403
##
## Mcnemar's Test P-Value : < 2.2e-16
##
## Statistics by Class:
##
##              Class: Canonical Class: Non-Canonical Class: Crying
## Sensitivity              0.7484              0.6686              0.31561
## Specificity              0.8916              0.7069              0.95269
## Pos Pred Value           0.6352              0.6910              0.51509
## Neg Pred Value           0.9336              0.6852              0.89735
## Prevalence               0.2014              0.4950              0.13737
## Detection Rate           0.1507              0.3310              0.04335
## Detection Prevalence     0.2373              0.4790              0.08417
## Balanced Accuracy        0.8200              0.6878              0.63415
##
##              Class: Laughing Class: Junk
## Sensitivity              0.311828          0.63066
## Specificity              0.991982          0.88834
## Pos Pred Value           0.542056          0.47217
## Neg Pred Value           0.979321          0.93822
## Prevalence               0.029538          0.13673
## Detection Rate           0.009211          0.08623
## Detection Prevalence     0.016992          0.18263
## Balanced Accuracy        0.651905          0.75950

pdf("../Results/precision_TD.pdf",height=10,width=10)
colsums=colSums(conf_tab)
my_conf_tab=conf_tab
for(i in 1:5) my_conf_tab[,i]=my_conf_tab[,i]/colsums[i]
colSums(my_conf_tab)

##      Canonical Non-Canonical          Crying          Laughing          Junk
##              1              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$Zooniverse)
colnames(prop_cat)[3]<-"pr"
data.frame(conf_tab)->stall
```



```

stall$id=paste(stall$Lab,stall$Zooniverse)
stall=merge(stall,prop_cat[c("id","pr")])
ggplot(data = stall, mapping = aes(y = Lab, x=Zooniverse)) +
  geom_tile(aes(fill= rescale(pr)), colour = "white") +
  geom_text(aes(label = paste(round(pr,"%")), vjust = -1,size=8) +
  geom_text(aes(label = Freq), vjust = 1,size=8) +
  scale_fill_gradient(low = "white", high = "red", name = "Proportion") +
  theme(legend.position = "none") +
  xlab("Zooniverse") + ylab("Lab") +
  ggtitle("Precision")+theme(text = element_text(size=20),
    axis.text.x = element_text(angle=90, hjust=1))
dev.off()

## pdf
## 2

pdf("../Results/recall_TD.pdf",height=10,width=10)
prop_cat=data.frame(conf_tab/rowSums(conf_tab)*100) #generates recall because rows
prop_cat$id=paste(prop_cat$Lab,prop_cat$Zooniverse)
colnames(prop_cat)[3]<-"rec"
data.frame(conf_tab)->stall
stall$id=paste(stall$Lab,stall$Zooniverse)
stall=merge(stall,prop_cat[c("id","rec")])
ggplot(data = stall, mapping = aes(y = Lab, x=Zooniverse)) +
  geom_tile(aes(fill= rescale(rec)), colour = "white") +
  geom_text(aes(label = paste(round(rec,"%")), vjust = -1,size=8) +
  geom_text(aes(label = Freq), vjust = 1,size=8) +
  scale_fill_gradient(low = "white", high = "red", name = "Proportion") +
  theme(legend.position = "none") +
  xlab("Zooniverse") + ylab("Lab") +
  ggtitle("Recall")+theme(text = element_text(size=20),
    axis.text.x = element_text(angle=90, hjust=1))
dev.off()

## pdf
## 2

```

Child level descriptors

Although there may be errors at the level of the segment, what we really care about is whether Zooniverse annotations give a reliable image of the child's individual development. This is what we look at in this section. In all of these graphs, red points correspond to children diagnosed with Angelman Syndrome, black for low-risk control.

```

#get the ns by child, then calculate the linguistic ratio & canonical ratio, separately for zooniverse
ztab=table(data_all$filename,data_all$Zoon_classif)
z_lr=rowSums(ztab[,c("Canonical","Non-Canonical")])/rowSums(ztab[,~which(colnames(ztab) %in% c("Junk"))])
z_cr=ztab[,c("Canonical")]/rowSums(ztab[,c("Canonical","Non-Canonical")])
ltab=table(data_all$filename,data_all$lab)
l_lr=rowSums(ltab[,c("Canonical","Non-Canonical")])/rowSums(ltab[,~which(colnames(ztab) %in% c("Junk"))])
l_cr=ltab[,c("Canonical")]/rowSums(ltab[,c("Canonical","Non-Canonical")])
#put all the ratios together
if(sum(rownames(ztab)==rownames(ltab))==dim(ztab)[1]) ratios=cbind(rownames(ztab),z_lr,z_cr,l_lr,l_cr)
colnames(ratios)[1]<-"filename"

```

```

#ages=aggregate(data_all$Age,by=list(data_all$ChildID),mean) #this is a weird way of adding ages, since
#improvement: now we merge with a demo data tab, but note this is merged with child id, so the problem

# Created demo_data with filenames. Use filenames instead of childIDs to merge ratios and demo data.

merge(ratios,demo_data_fn,by="filename")->ratios
colnames(ratios)[dim(ratios)[2]]<-"Age"
#cbinding results in text, so we numerize the ratios
for(thisvar in c("z_lr","z_cr","l_lr","l_cr")) ratios[,thisvar]=as.numeric(as.character(ratios[,thisvar]))
summary(ratios)

```

```

##          filename      z_lr      z_cr
## 20180206_110905_009463: 1   Min.    :0.5927   Min.    :0.02882
## 20180419_111712_022875: 1   1st Qu.:0.8058   1st Qu.:0.05891
## 20180530_180405_024879: 1   Median :0.8785   Median :0.14065
## 20180530_181101_022875: 1   Mean    :0.8511   Mean    :0.19128
## 20180808_111325_024882: 1   3rd Qu.:0.9313   3rd Qu.:0.28454
## 20180906_133011_022875: 1   Max.    :0.9740   Max.    :0.51084
## (Other)                  :14
##      l_lr      l_cr      ChildID      Diagnosis
## Min.    :0.3880   Min.    :0.01488   Length:20   AngelmanSyndrome:10
## 1st Qu.:0.9130   1st Qu.:0.06609   Class :character   Low-RiskControl :10
## Median :0.9578   Median :0.11115   Mode  :character
## Mean    :0.9131   Mean    :0.19129
## 3rd Qu.:0.9787   3rd Qu.:0.26529
## Max.    :0.9968   Max.    :0.60000
##
## Sex      Age
## F: 7     Min.    : 4.67
## M:13     1st Qu.:12.13
##          Median :17.50
##          Mean    :25.04
##          3rd Qu.:43.46
##          Max.    :53.26
##

```

Correlations with age

We first look generally at two measures that have been found to relate to age:

- linguistic ratio = (“Canonical”+“Non-Canonical”)/“All vocalizations” (i.e. we remove junk)
- canonical ratio = “Canonical”/(“Canonical”+“Non-Canonical”) (i.e. we remove junk + non-linguistic vocalizations)

TODO

- make margins smaller
- remove title repetition (age only at the bottom)

```

prettynames=c("Linguistic Ratio (Zooniverse)","Canonical Ratio (Zooniverse)",
              "Linguistic Ratio (Lab)","Canonical Ratio (Lab)" )
names(prettynames)<-c("z_lr","z_cr","l_lr","l_cr")
mycols=c("black","red")
names(mycols)<-c("Low-RiskControl","AngelmanSyndrome")

```

```

mypch=c(4,20)
names(mypch)<-c("Low-RiskControl","AngelmanSyndrome")

jpeg("../Results/corage.jpg",width=20,height=20,units="cm",res=300)
layout(matrix(c(1:4), 2, 2, byrow = F))
for(thisvar in c("z_lr","z_cr","l_lr","l_cr")) {

  plot(ratios[,thisvar]~ratios$Age, pch=mypch[ratios$Diagnosis],xlab="Age (months)",ylab=prettynames[thisvar],
       col=mycols[ratios$Diagnosis])
  abline(lm(ratios[,thisvar]~ratios$Age,subset=c(ratios$Diagnosis=="AngelmanSyndrome")),col="black")
  myr=paste0("r=",round(cor.test(ratios[ratios$Diagnosis=="AngelmanSyndrome",thisvar],ratios$Age[ratios$Diagnosis=="AngelmanSyndrome"],method="s",p.adjust="none"))$estimate,2))
  text(mean(ratios$Age[ratios$Diagnosis=="AngelmanSyndrome"]),mean(ratios[ratios$Diagnosis=="AngelmanSyndrome",thisvar]),myr,col="black")
  abline(lm(ratios[,thisvar]~ratios$Age,subset=c(ratios$Diagnosis!="AngelmanSyndrome")),col="red")
  myr=paste0("r=",round(cor.test(ratios[ratios$Diagnosis!="AngelmanSyndrome",thisvar],ratios$Age[ratios$Diagnosis!="AngelmanSyndrome"],method="s",p.adjust="none"))$estimate,2))
  text(mean(ratios$Age[ratios$Diagnosis!="AngelmanSyndrome"]),mean(ratios[ratios$Diagnosis!="AngelmanSyndrome",thisvar]),myr,col="red")
}
dev.off()

## pdf
## 2

```

Correlations across zooniverse and lab

But the key thing for us: Are Zooniverse annotations describing children similar to lab annotations? The answer is clearly yes.

TODO:

- add code to print out the results paragraphs
- change figure generation to have lines fitted to each group

```

#Ling ratio
pdf("../Results/ling_rat_z_vs_l_final.pdf",height=5,width=5)
lims=range(c(ratios[, "z_lr"],ratios[, "l_lr"]))
myr=round(cor.test(ratios[, "z_lr"],ratios[, "l_lr"])$estimate,3)
plot(ratios[, "z_lr"]~ratios[, "l_lr"], pch=20,xlab=prettynames["l_lr"],ylab=prettynames["z_lr"],main=paste0("Ling ratio",myr))
xlim=lims,ylim=lims,
col=mycols[ratios$Diagnosis])
abline(lm(ratios[, "z_lr"]~ratios[, "l_lr"]))
lines(c(0,1),c(0,1),lty=2,col="darkgray")
dev.off()

## pdf
## 2

```

```

#CR
pdf("../Results/can_rat_z_vs_l_final.pdf",height=5,width=5)
lims=range(c(ratios[, "z_cr"],ratios[, "l_cr"]))
myr=round(cor.test(ratios[, "z_cr"],ratios[, "l_cr"])$estimate,3)
plot(ratios[, "z_cr"]~ratios[, "l_cr"], pch=20,xlab=prettynames["l_cr"],ylab=prettynames["z_cr"],main=paste0("CR",myr))
xlim=lims,ylim=lims,
col=mycols[ratios$Diagnosis])
abline(lm(ratios[, "z_cr"]~ratios[, "l_cr"]),col="darkgray")
lines(c(0,1),c(0,1),lty=2,col="darkgray")
dev.off()

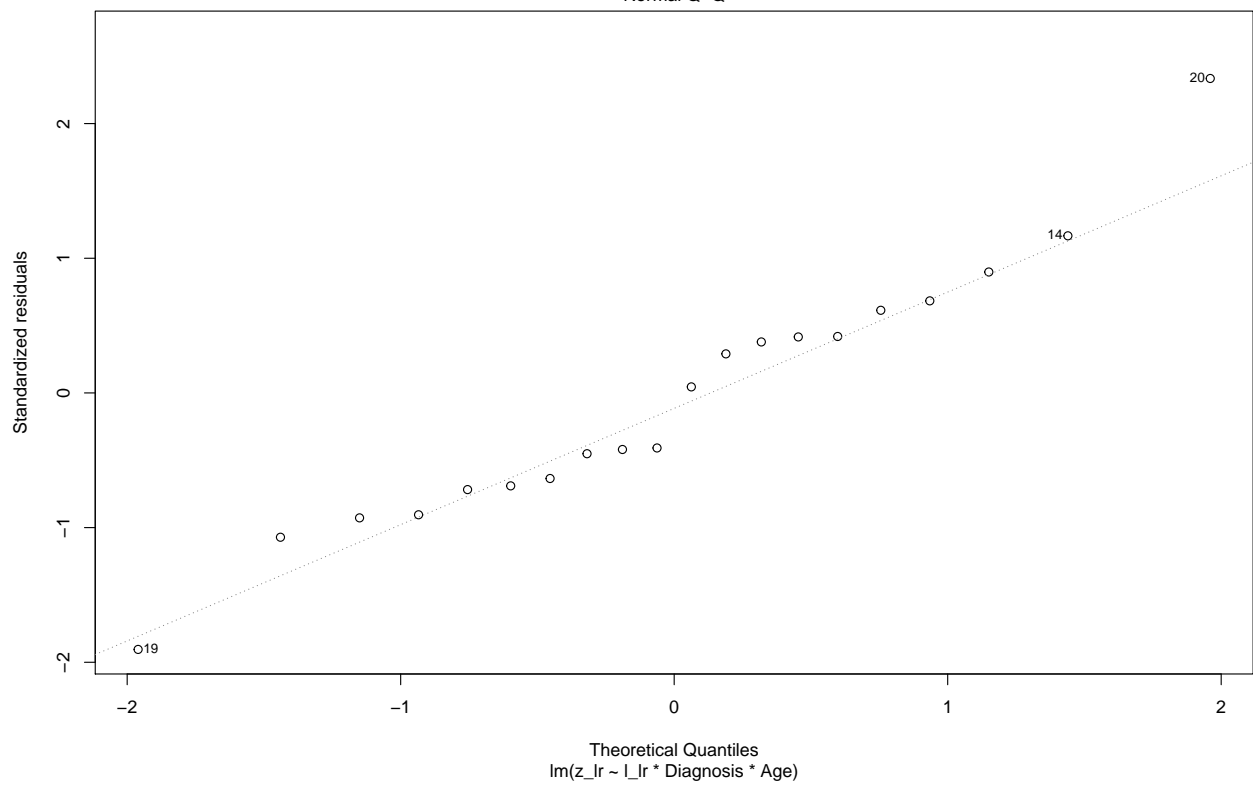
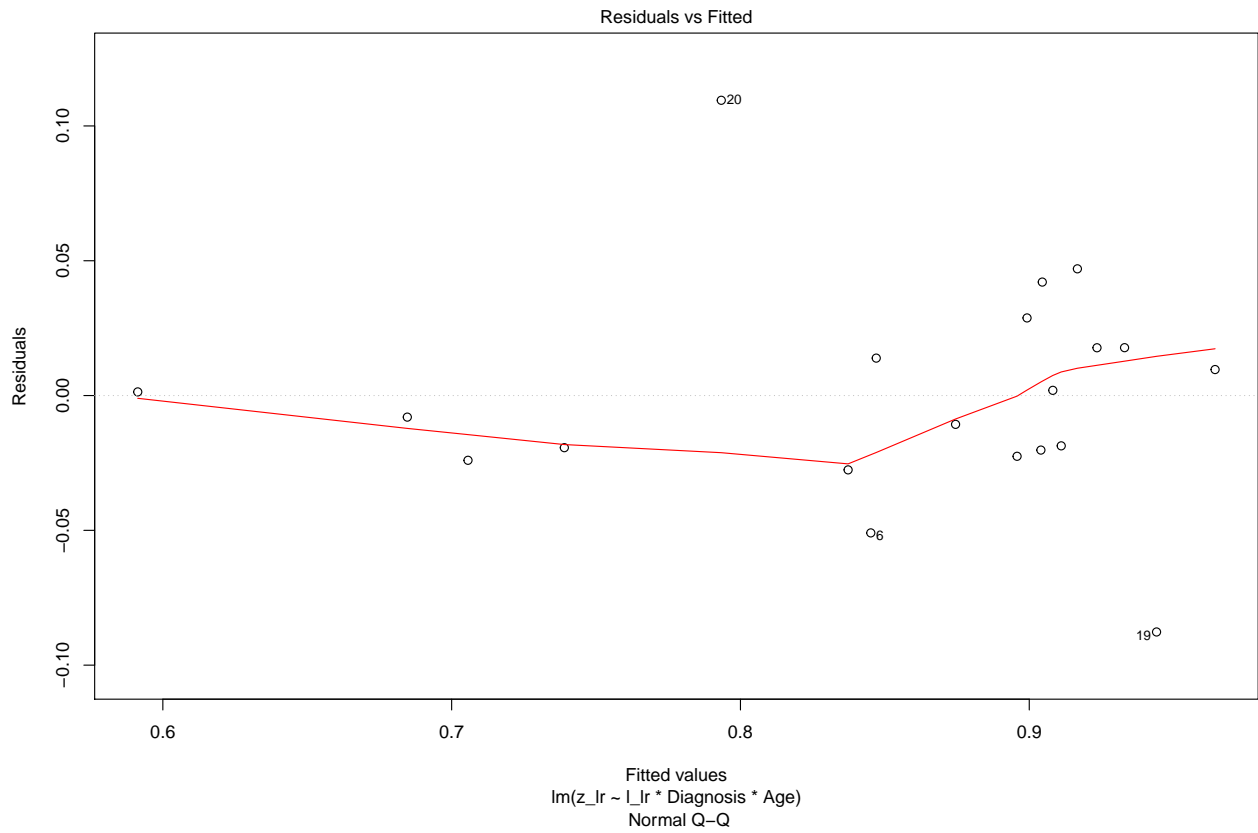
```

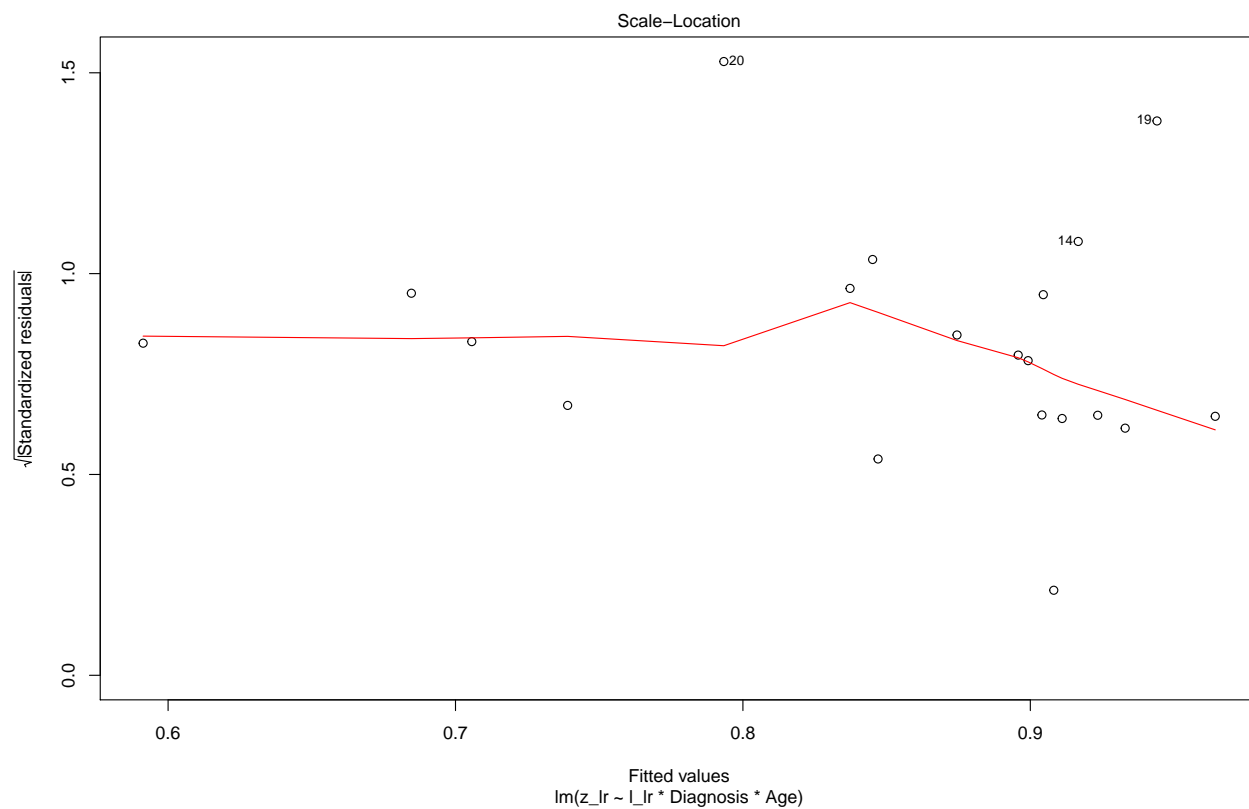
```
## pdf
## 2

lin_mod=lm(z_lr~l_lr*Diagnosis*Age,data=ratios)
summary(lin_mod)

##
## Call:
## lm(formula = z_lr ~ l_lr * Diagnosis * Age, data = ratios)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.087697 -0.020805 -0.003327  0.017732  0.109518
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -1.89796     2.53800   -0.748   0.4690
## l_lr           2.95125     2.65350    1.112   0.2878
## DiagnosisLow-RiskControl 2.84516     2.54890    1.116   0.2862
## Age            0.03583     0.05740    0.624   0.5442
## l_lr:DiagnosisLow-RiskControl -3.03307     2.66494   -1.138   0.2773
## l_lr:Age       -0.03819     0.05996   -0.637   0.5362
## DiagnosisLow-RiskControl:Age -0.15256     0.07332   -2.081   0.0595 .
## l_lr:DiagnosisLow-RiskControl:Age 0.16017     0.07535    2.126   0.0550 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.05096 on 12 degrees of freedom
## Multiple R-squared:  0.8575, Adjusted R-squared:  0.7744
## F-statistic: 10.32 on 7 and 12 DF,  p-value: 0.0003014

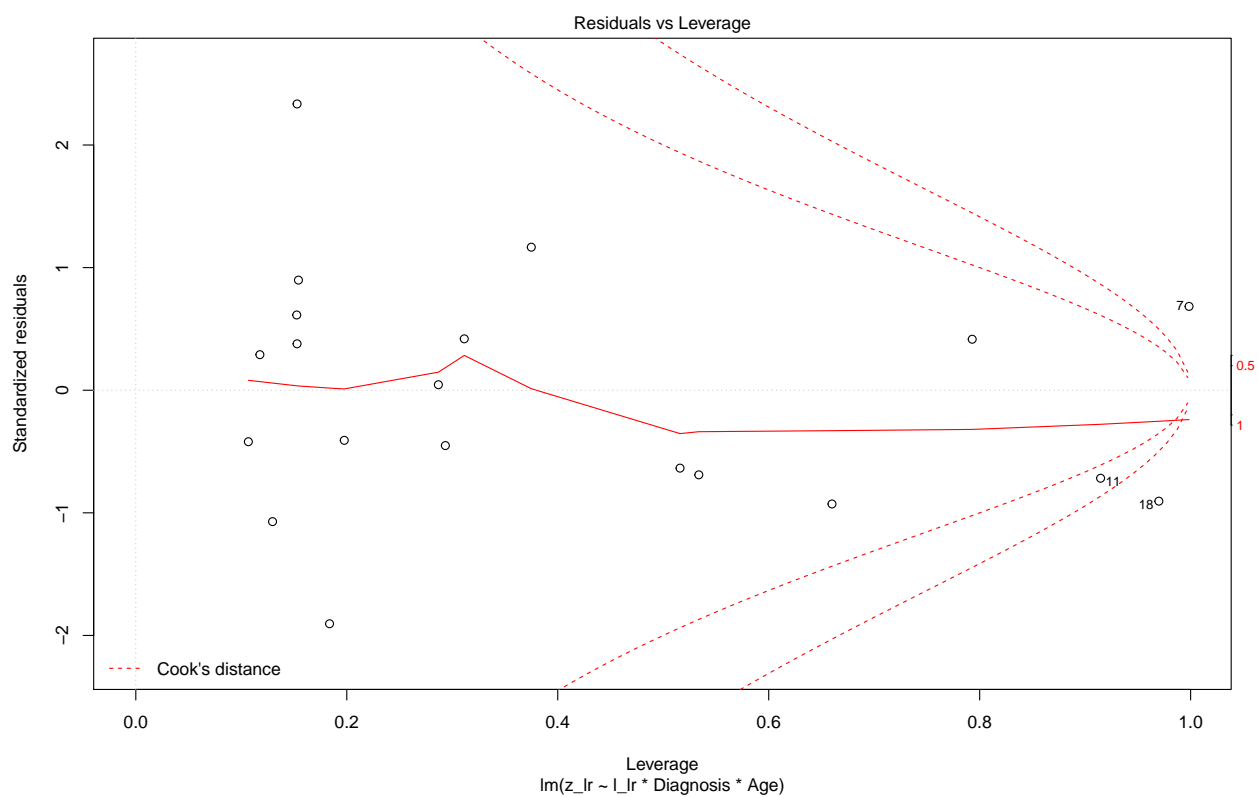
plot(lin_mod)
```





Warning in sqrt(crit * p * (1 - hh)/hh): NaNs produced

Warning in sqrt(crit * p * (1 - hh)/hh): NaNs produced



Code junkyard

This is redundant code

Different pdfs for diff groups

```
prettynames=c("AS Linguistic Ratio (Zooniverse)","AS Canonical Ratio (Zooniverse)",
              "AS Linguistic Ratio (Lab)","AS Canonical Ratio (Lab)" )
names(prettynames)<-c("z_lr","z_cr","l_lr","l_cr")

AS_ratios<-subset(ratios, Diagnosis=="AngelmanSyndrome")
TD_ratios<-subset(ratios, Diagnosis=="Low-RiskControl")

for(thisvar in c("z_lr","z_cr","l_lr","l_cr")) {
  myr=round(cor.test(AS_ratios[,thisvar],AS_ratios$Age)$estimate,3)
  plot(AS_ratios[,thisvar]~AS_ratios$Age, pch=20,xlab="Age (months)",ylab=prettynames[thisvar],main=pas)
  abline(lm(AS_ratios[,thisvar]~AS_ratios$Age))
}

prettynames=c("TD Linguistic Ratio (Zooniverse)","TD Canonical Ratio (Zooniverse)",
              "TD Linguistic Ratio (Lab)","TD Canonical Ratio (Lab)" )
for(thisvar in c("z_lr","z_cr","l_lr","l_cr")) {
  myr=round(cor.test(TD_ratios[,thisvar],TD_ratios$Age)$estimate,3)
  plot(TD_ratios[,thisvar]~TD_ratios$Age, pch=20,xlab="Age (months)",ylab=prettynames[thisvar],main=pas)
  abline(lm(TD_ratios[,thisvar]~TD_ratios$Age))
}

#Ling ratio
pdf("../Results/td_ling_rat_z_vs_l_final.pdf",height=5,width=5)
lims=range(c(TD_ratios[, "z_lr"],TD_ratios[, "l_lr"]))
myr=round(cor.test(TD_ratios[, "z_lr"],TD_ratios[, "l_lr"])$estimate,3)
plot(TD_ratios[, "z_lr"]~TD_ratios[, "l_lr"], pch=20,xlab=prettynames["l_lr"],ylab=prettynames["z_lr"],
     xlim=lims,ylim=lims,
     col=mycols[TD_ratios$Diagnosis])
abline(lm(TD_ratios[, "z_lr"]~TD_ratios[, "l_lr"]))
lines(c(0,1),c(0,1),lty=2,col="darkgray")
dev.off()

#CR
pdf("../Results/td_can_rat_z_vs_l_final.pdf",height=5,width=5)
lims=range(c(TD_ratios[, "z_cr"],TD_ratios[, "l_cr"]))
myr=round(cor.test(TD_ratios[, "z_cr"],TD_ratios[, "l_cr"])$estimate,3)
plot(TD_ratios[, "z_cr"]~TD_ratios[, "l_cr"], pch=20,xlab=prettynames["l_cr"],ylab=prettynames["z_cr"],
     xlim=lims,ylim=lims,
     col=mycols[TD_ratios$Diagnosis])
abline(lm(TD_ratios[, "z_cr"]~TD_ratios[, "l_cr"]),col="darkgray")
lines(c(0,1),c(0,1),lty=2,col="darkgray")
dev.off()

#Ling ratio
pdf("../Results/AS_ling_rat_z_vs_l_final.pdf",height=5,width=5)
lims=range(c(AS_ratios[, "z_lr"],AS_ratios[, "l_lr"]))
myr=round(cor.test(AS_ratios[, "z_lr"],AS_ratios[, "l_lr"])$estimate,3)
plot(AS_ratios[, "z_lr"]~AS_ratios[, "l_lr"], pch=20,xlab=prettynames["l_lr"],ylab=prettynames["z_lr"],
     xlim=lims,ylim=lims,
     col=mycols[AS_ratios$Diagnosis])
abline(lm(AS_ratios[, "z_lr"]~AS_ratios[, "l_lr"]))
```

```

    lines(c(0,1),c(0,1),lty=2,col="darkgray")
dev.off()
#CR
pdf("../Results/AS_can_rat_z_vs_l_final.pdf",height=5,width=5)
lims=range(c(AS_ratios[, "z_cr"],AS_ratios[, "l_cr"]))
myr=round(cor.test(AS_ratios[, "z_cr"],AS_ratios[, "l_cr"])$estimate,3)
plot(AS_ratios[, "z_cr"]~AS_ratios[, "l_cr"], pch=20,xlab=prettynames["l_cr"],ylab=prettynames["z_cr"],
     xlim=lims,ylim=lims,
     col=mycols[AS_ratios$Diagnosis])
abline(lm(AS_ratios[, "z_cr"]~AS_ratios[, "l_cr"]),col="darkgray")
lines(c(0,1),c(0,1),lty=2,col="darkgray")
dev.off()

```

combined pdf

```

#COMBINED to save space
pdf("../Results/combined_final.pdf",height=5,width=5)
lims=range(c(ratios[, "z_lr"],ratios[, "l_lr"]),c(ratios[, "z_cr"],ratios[, "l_cr"]))
#myr=round(cor.test(ratios[, "z_lr"],ratios[, "l_lr"])$estimate,3)
plot(ratios[, "z_lr"]~ratios[, "l_lr"],xlab="Laboratory annotations",ylab="Zooniverse annotations",
     xlim=lims,ylim=lims,
     pch=20,col=mycols[ratios$Diagnosis])
points(ratios[, "z_cr"]~ratios[, "l_cr"], pch=2, col=mycols[ratios$Diagnosis])
abline(lm(ratios[, "z_cr"]~ratios[, "l_cr"]))
abline(lm(ratios[, "z_lr"]~ratios[, "l_lr"]),lty=3)
# lines(c(0,1),c(0,1),lty=2,col="darkgray")
dev.off()

```