

Analysis-Brain-AllData

Setup

Enviroment Setup

```
#setwd('~/Desktop/andlab/code')  
odf <- read.csv('./brain_cb.csv')  
odf <- odf[complete.cases(odf), ]  
  
library('caret')
```

```
## Loading required package: lattice
```

```
## Loading required package: ggplot2
```

```
#setwd('~/Desktop/andlab/code')
```

Function Define

```

process <- function(df, reg, upr, plot=FALSE){

  newdf <- data.frame('y' = df$aggressive_sumscore)
  newdf$fitted <- reg$fitted.values

  newdf$class <- ifelse(newdf$y <= (-2.02065), 'L', ifelse(newdf$y>(upr),'M','H'))
  #change == -2.020650971 to change <= -2.020650971 to solve prediction for 0 is 0
  newdf$pred_class <-ifelse(newdf$fitted <= (-2.02065), 'L', ifelse(newdf$fitted>(up
r),'M','H'))

  print(mean(newdf$pred_class == newdf$class))
  #cm <- confusionMatrix(factor(newdf$pred_class, levels = 0:2), factor(newdf$class,
levels = 0:2) )
  cm <- confusionMatrix(table(newdf$class,newdf$pred_class))
  if(plot)
  {
    par(mfrow = c(2, 2))
    plot(reg)
  }
  return(cm=cm)
}

process_2_level <- function(df, reg, plot=FALSE){

  newdf <- data.frame('y' = df$aggressive_sumscore)
  newdf$fitted <- reg$fitted.values

  newdf$class <- ifelse(newdf$y <= (-2.020650971), 'L', 'H')
  #change == -2.020650971 to change <= -2.020650971 to solve prediction for 0 is 0
  newdf$pred_class <-ifelse(newdf$fitted <= (-2.020650971), 'L', 'H')

  print(mean(newdf$pred_class == newdf$class))
  #cm <- confusionMatrix(factor(newdf$pred_class, levels = 0:2), factor(newdf$class,
levels = 0:2) )
  cm <- confusionMatrix(table(newdf$class,newdf$pred_class))
  if(plot)
  {
    par(mfrow = c(2, 2))
    plot(reg)
  }
  return(cm=cm)
}

```

Analysis W/ All Data, All Attributes

Attribute selection

```
df <- subset(odf, select==c(prosocial_child, prosocial_parent, subjectkey))
```

OLS

Model Selection

```
reg <- lm(df$aggressive_sumscore ~ ., data=df)
```

Result

```
summary(reg)
```

```
##
## Call:
## lm(formula = df$aggressive_sumscore ~ ., data = df)
##
## Residuals:
```

| | Min | 1Q | Median | 3Q | Max |
|--|--------|--------|--------|-------|--------|
| | -3.644 | -1.914 | -1.471 | 0.386 | 32.773 |

```
##
## Coefficients:
```

| | Estimate | Std. Error | t value | Pr(> t) | |
|-----------------------------|-----------|------------|---------|----------|----|
| (Intercept) | 0.224087 | 0.089894 | 2.493 | 0.01269 | * |
| rsfmri_cor_ngd_cerc_scs_thp | -0.584585 | 0.285513 | -2.047 | 0.04063 | * |
| rsfmri_cor_ngd_cerc_scs_cde | -0.053300 | 0.345053 | -0.154 | 0.87724 | |
| rsfmri_cor_ngd_cerc_scs_pt | -0.426288 | 0.301887 | -1.412 | 0.15795 | |
| rsfmri_cor_ngd_cerc_scs_hp | -0.483225 | 0.305502 | -1.582 | 0.11374 | |
| rsfmri_cor_ngd_cerc_scs_ag | -0.695918 | 0.216092 | -3.220 | 0.00128 | ** |
| rsfmri_cor_ngd_cerc_scs_aa | 0.310640 | 0.238658 | 1.302 | 0.19308 | |
| rsfmri_cor_ngd_cerc_scs_bs | 0.008034 | 0.238067 | 0.034 | 0.97308 | |
| rsfmri_cor_ngd_df_scs_thp | -0.124255 | 0.440536 | -0.282 | 0.77791 | |
| rsfmri_cor_ngd_df_scs_cde | 0.309316 | 0.288583 | 1.072 | 0.28381 | |
| rsfmri_cor_ngd_df_scs_pt | 0.367106 | 0.346409 | 1.060 | 0.28928 | |
| rsfmri_cor_ngd_df_scs_hp | -0.559522 | 0.281512 | -1.988 | 0.04688 | * |
| rsfmri_cor_ngd_df_scs_ag | 0.118085 | 0.310568 | 0.380 | 0.70379 | |
| rsfmri_cor_ngd_df_scs_aa | -0.133659 | 0.199450 | -0.670 | 0.50279 | |
| rsfmri_cor_ngd_df_scs_bs | -0.085437 | 0.153257 | -0.557 | 0.57722 | |
| rsfmri_cor_ngd_dsa_scs_thp | -0.361705 | 0.266768 | -1.356 | 0.17516 | |
| rsfmri_cor_ngd_dsa_scs_cde | -0.175751 | 0.301879 | -0.582 | 0.56045 | |
| rsfmri_cor_ngd_dsa_scs_pt | 0.153157 | 0.382552 | 0.400 | 0.68890 | |
| rsfmri_cor_ngd_dsa_scs_hp | -0.132386 | 0.412192 | -0.321 | 0.74808 | |
| rsfmri_cor_ngd_dsa_scs_ag | 0.912593 | 0.435551 | 2.095 | 0.03617 | * |
| rsfmri_cor_ngd_dsa_scs_aa | -0.293120 | 0.294888 | -0.994 | 0.32024 | |
| rsfmri_cor_ngd_dsa_scs_bs | -0.570311 | 0.275608 | -2.069 | 0.03854 | * |
| rsfmri_cor_ngd_fopa_scs_thp | 0.144710 | 0.254378 | 0.569 | 0.56945 | |
| rsfmri_cor_ngd_fopa_scs_cde | 0.072505 | 0.454776 | 0.159 | 0.87333 | |
| rsfmri_cor_ngd_fopa_scs_pt | -0.309079 | 0.301522 | -1.025 | 0.30536 | |
| rsfmri_cor_ngd_fopa_scs_hp | -0.245102 | 0.294756 | -0.832 | 0.40569 | |
| rsfmri_cor_ngd_fopa_scs_ag | -0.396177 | 0.287707 | -1.377 | 0.16854 | |
| rsfmri_cor_ngd_fopa_scs_aa | 0.053215 | 0.330627 | 0.161 | 0.87213 | |
| rsfmri_cor_ngd_fopa_scs_bs | 0.035860 | 0.177372 | 0.202 | 0.83979 | |
| rsfmri_cor_ngd_sa_scs_thp | 0.000434 | 0.279149 | 0.002 | 0.99876 | |
| rsfmri_cor_ngd_sa_scs_cde | -0.048681 | 0.229523 | -0.212 | 0.83204 | |
| rsfmri_cor_ngd_sa_scs_pt | -0.700347 | 0.343962 | -2.036 | 0.04176 | * |
| rsfmri_cor_ngd_sa_scs_hp | 0.562652 | 0.262898 | 2.140 | 0.03236 | * |
| rsfmri_cor_ngd_sa_scs_ag | 0.204990 | 0.284836 | 0.720 | 0.47174 | |
| rsfmri_cor_ngd_sa_scs_aa | 0.029209 | 0.454840 | 0.064 | 0.94880 | |
| rsfmri_cor_ngd_sa_scs_bs | 0.162003 | 0.238182 | 0.680 | 0.49641 | |
| rsfmri_cor_ngd_vta_scs_thp | -0.113448 | 0.248408 | -0.457 | 0.64790 | |
| rsfmri_cor_ngd_vta_scs_cde | -0.285884 | 0.359014 | -0.796 | 0.42587 | |
| rsfmri_cor_ngd_vta_scs_pt | -0.049497 | 0.371129 | -0.133 | 0.89391 | |
| rsfmri_cor_ngd_vta_scs_hp | -0.502686 | 0.304682 | -1.650 | 0.09900 | . |
| rsfmri_cor_ngd_vta_scs_ag | -0.621755 | 0.369771 | -1.681 | 0.09270 | . |

```
## rsfmri_cor_ngd_vta_scs_aa    0.259894    0.280487    0.927    0.35417
## rsfmri_cor_ngd_vta_scs_bs    0.052903    0.410540    0.129    0.89747
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.479 on 11041 degrees of freedom
## Multiple R-squared:  0.008874,    Adjusted R-squared:  0.005104
## F-statistic: 2.354 on 42 and 11041 DF,  p-value: 1.894e-06
```

```
cm <- process(df, reg, -0.73254990)
```

```
## [1] 0.345904
```

```
print(cm)
```

```
## Confusion Matrix and Statistics
##
##
##      H      L      M
## H   46      1 2852
## L   57      2 4285
## M   50      5 3786
##
## Overall Statistics
##
##              Accuracy : 0.3459
##              95% CI   : (0.337, 0.3548)
##    No Information Rate : 0.9855
##    P-Value [Acc > NIR] : 1
##
##              Kappa   : 8e-04
##
## Mcnemar's Test P-Value : <2e-16
##
## Statistics by Class:
##
##              Class: H  Class: L  Class: M
## Sensitivity          0.30065 0.2500000  0.34661
## Specificity          0.73900 0.6079812  0.65839
## Pos Pred Value       0.01587 0.0004604  0.98568
## Neg Pred Value       0.98693 0.9991098  0.01463
## Prevalence           0.01380 0.0007218  0.98547
## Detection Rate       0.00415 0.0001804  0.34157
## Detection Prevalence 0.26155 0.3919163  0.34654
## Balanced Accuracy     0.51983 0.4289906  0.50250
```

```
print(cm$byClass)
```

```
##          Sensitivity Specificity Pos Pred Value Neg Pred Value    Precision
## Class: H    0.3006536    0.7389992    0.0158675405    0.98692731 0.0158675405
## Class: L    0.2500000    0.6079812    0.0004604052    0.99910979 0.0004604052
## Class: M    0.3466081    0.6583851    0.9856808123    0.01463482 0.9856808123
##          Recall          F1    Prevalence Detection Rate
## Class: H 0.3006536 0.0301441678 0.0138036810    0.0041501263
## Class: L 0.2500000 0.0009191176 0.0007217611    0.0001804403
## Class: M 0.3466081 0.5128691412 0.9854745579    0.3415734392
##          Detection Prevalence Balanced Accuracy
## Class: H          0.2615482          0.5198264
## Class: L          0.3919163          0.4289906
## Class: M          0.3465355          0.5024966
```

OLS W/ Interaction

Model Selection

```
reg <- lm(df$aggressive_sumscore ~ . + .^2, data=df)
```

Result

```
#summary(reg)
cm <- process(df, reg, -0.73254990)
```

```
## [1] 0.3612414
```

```
print(cm)
```

```
## Confusion Matrix and Statistics
##
##
##      H      L      M
## H  393    92 2414
## L  561   165 3618
## M  357    38 3446
##
## Overall Statistics
##
##              Accuracy : 0.3612
##              95% CI : (0.3523, 0.3703)
##      No Information Rate : 0.8551
##      P-Value [Acc > NIR] : 1
##
##              Kappa : 0.0356
##
##  Mcnemar's Test P-Value : <2e-16
##
## Statistics by Class:
##
##              Class: H Class: L Class: M
## Sensitivity      0.29977  0.55932  0.3636
## Specificity      0.74358  0.61266  0.7540
## Pos Pred Value   0.13556  0.03798  0.8972
## Neg Pred Value   0.88784  0.98071  0.1672
## Prevalence       0.11828  0.02661  0.8551
## Detection Rate   0.03546  0.01489  0.3109
## Detection Prevalence 0.26155  0.39192  0.3465
## Balanced Accuracy 0.52168  0.58599  0.5588
```

```
print(cm$byClass)
```

```
##      Sensitivity Specificity Pos Pred Value Neg Pred Value Precision
## Class: H  0.2997712  0.7435792  0.13556399  0.8878436 0.13556399
## Class: L  0.5593220  0.6126610  0.03798343  0.9807122 0.03798343
## Class: M  0.3635788  0.7540473  0.89716220  0.1671959 0.89716220
##      Recall      F1 Prevalence Detection Rate Detection Prevalence
## Class: H 0.2997712 0.18669834 0.11827860  0.03545651  0.2615482
## Class: L 0.5593220 0.07113602 0.02661494  0.01488632  0.3919163
## Class: M 0.3635788 0.51745627 0.85510646  0.31089859  0.3465355
##      Balanced Accuracy
## Class: H      0.5216752
## Class: L      0.5859915
## Class: M      0.5588131
```

Stepwise

Model Selection

```

null <- lm(aggressive_sumscore ~ 1, data=df)
full <- lm(aggressive_sumscore ~ ., data=df)
reg <- step(null, scope=formula(full), direction="forward", k=log(nrow(df)), trace=0)

```

Result

```
summary(reg)
```

```

##
## Call:
## lm(formula = aggressive_sumscore ~ rsfmri_cor_ngd_cerc_scs_ag +
##     rsfmri_cor_ngd_vta_scs_hp, data = df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.532 -1.943 -1.514  0.376 33.418
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      0.05564    0.04032   1.380 0.167681
## rsfmri_cor_ngd_cerc_scs_ag -0.85628    0.18915  -4.527 6.04e-06 ***
## rsfmri_cor_ngd_vta_scs_hp -0.86647    0.26138  -3.315 0.000919 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.483 on 11081 degrees of freedom
## Multiple R-squared:  0.002821, Adjusted R-squared:  0.002641
## F-statistic: 15.68 on 2 and 11081 DF, p-value: 1.59e-07

```

```
cm <- process(df, reg, -0.73254990)
```

```
## [1] 0.3447311
```

```
print(cm)
```



```
## Confusion Matrix and Statistics
##
##
##           H      L      M
## H      5      0 2894
## L      9      0 4335
## M     24      1 3816
##
## Overall Statistics
##
##           Accuracy : 0.3447
##           95% CI : (0.3359, 0.3537)
##       No Information Rate : 0.9965
##       P-Value [Acc > NIR] : 1
##
##           Kappa : -0.0023
##
## Mcnemar's Test P-Value : <2e-16
##
## Statistics by Class:
##
##           Class: H  Class: L Class: M
## Sensitivity      0.1315789 0.000e+00 0.345496
## Specificity      0.7380047 6.080e-01 0.358974
## Pos Pred Value   0.0017247 0.000e+00 0.993491
## Neg Pred Value   0.9959682 9.999e-01 0.001933
## Prevalence       0.0034284 9.022e-05 0.996481
## Detection Rate   0.0004511 0.000e+00 0.344280
## Detection Prevalence 0.2615482 3.919e-01 0.346536
## Balanced Accuracy 0.4347918 3.040e-01 0.352235
```

```
print(cm$byClass)
```

```
##           Sensitivity Specificity Pos Pred Value Neg Pred Value Precision
## Class: H  0.1315789  0.7380047  0.001724733  0.995968235 0.001724733
## Class: L  0.0000000  0.6080484  0.000000000  0.999851632 0.000000000
## Class: M  0.3454957  0.3589744  0.993491278  0.001932901 0.993491278
##           Recall      F1    Prevalence Detection Rate Detection Prevalence
## Class: H 0.1315789 0.003404835 3.428365e-03 0.0004511007 0.2615482
## Class: L 0.0000000      NaN 9.022014e-05 0.0000000000 0.3919163
## Class: M 0.3454957 0.512696493 9.964814e-01 0.3442800433 0.3465355
##           Balanced Accuracy
## Class: H 0.4347918
## Class: L 0.3040242
## Class: M 0.3522350
```

Stepwise W/ Interaction

Model Selection

```

null <- lm(aggressive_sumscore ~ 1, data=df)
full <- lm(aggressive_sumscore ~ . + .^2, data=df)
reg <- step(null, scope=formula(full), direction="forward", k=log(nrow(df)), trace=0)

```

Result

```
summary(reg)
```

```

##
## Call:
## lm(formula = aggressive_sumscore ~ rsfmri_cor_ngd_cerc_scs_ag +
##     rsfmri_cor_ngd_vta_scs_hp, data = df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.532 -1.943 -1.514  0.376 33.418
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      0.05564    0.04032   1.380 0.167681
## rsfmri_cor_ngd_cerc_scs_ag -0.85628    0.18915  -4.527 6.04e-06 ***
## rsfmri_cor_ngd_vta_scs_hp  -0.86647    0.26138  -3.315 0.000919 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.483 on 11081 degrees of freedom
## Multiple R-squared:  0.002821, Adjusted R-squared:  0.002641
## F-statistic: 15.68 on 2 and 11081 DF, p-value: 1.59e-07

```

```
cm <- process(df, reg, -0.73254990)
```

```
## [1] 0.3447311
```

```
print(cm)
```

```
## Confusion Matrix and Statistics
##
##
##      H      L      M
## H      5      0 2894
## L      9      0 4335
## M     24      1 3816
##
## Overall Statistics
##
##              Accuracy : 0.3447
##              95% CI : (0.3359, 0.3537)
##      No Information Rate : 0.9965
##      P-Value [Acc > NIR] : 1
##
##              Kappa : -0.0023
##
##  Mcnemar's Test P-Value : <2e-16
##
## Statistics by Class:
##
##              Class: H  Class: L Class: M
## Sensitivity      0.1315789 0.000e+00 0.345496
## Specificity      0.7380047 6.080e-01 0.358974
## Pos Pred Value   0.0017247 0.000e+00 0.993491
## Neg Pred Value   0.9959682 9.999e-01 0.001933
## Prevalence       0.0034284 9.022e-05 0.996481
## Detection Rate   0.0004511 0.000e+00 0.344280
## Detection Prevalence 0.2615482 3.919e-01 0.346536
## Balanced Accuracy 0.4347918 3.040e-01 0.352235
```

```
print(cm$byClass)
```

```
##      Sensitivity Specificity Pos Pred Value Neg Pred Value Precision
## Class: H 0.1315789 0.7380047 0.001724733 0.995968235 0.001724733
## Class: L 0.0000000 0.6080484 0.000000000 0.999851632 0.000000000
## Class: M 0.3454957 0.3589744 0.993491278 0.001932901 0.993491278
##      Recall      F1      Prevalence Detection Rate Detection Prevalence
## Class: H 0.1315789 0.003404835 3.428365e-03 0.0004511007 0.2615482
## Class: L 0.0000000      NaN 9.022014e-05 0.0000000000 0.3919163
## Class: M 0.3454957 0.512696493 9.964814e-01 0.3442800433 0.3465355
##      Balanced Accuracy
## Class: H 0.4347918
## Class: L 0.3040242
## Class: M 0.3522350
```