# ARP2/3 complex associates with peroxisomes to participate in pexophagy in plants

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#### Statistical analysis

Full R source code written by Stanislav Vosolsobě

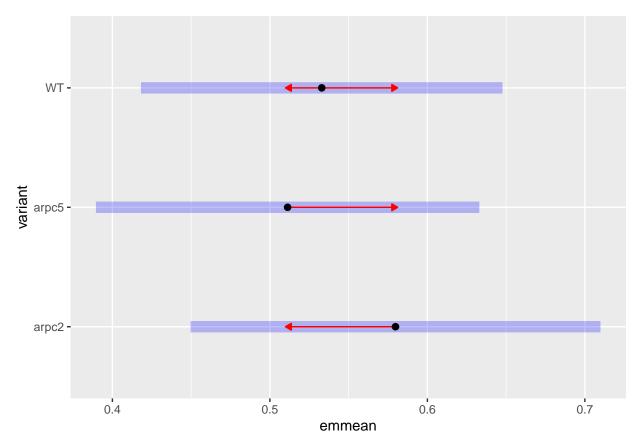
#### Required libraries

```
library(betareg) # for beta regression
library(lmtest) # for LR test after BetaReg
## Loading required package: zoo
##
## Attaching package: 'zoo'
## The following objects are masked from 'package:base':
##
##
      as.Date, as.Date.numeric
              # for zero-inflated Poisson model
library(pscl)
## Classes and Methods for R developed in the
## Political Science Computational Laboratory
## Department of Political Science
## Stanford University
## Simon Jackman
## hurdle and zeroinfl functions by Achim Zeileis
library(emmeans) # for multiple comparison
```

## Figure 2b

## \$emmeans

```
SE df lower.CL upper.CL
##
   variant emmean
##
    arpc2
            0.580 0.0646 45
                               0.450
                                        0.710
                               0.390
                                        0.633
##
    arpc5
             0.511 0.0604 45
             0.533 0.0570 45
                               0.418
                                        0.648
##
##
## Confidence level used: 0.95
##
## $contrasts
##
    contrast
                 estimate
                               SE df t.ratio p.value
  arpc2 - arpc5 0.0685 0.0885 45
                                      0.774 0.7206
##
## arpc2 - WT
                   0.0468 0.0862 45
                                      0.544 0.8502
  arpc5 - WT
                   -0.0217 0.0831 45 -0.261 0.9632
##
##
## P value adjustment: tukey method for comparing a family of 3 estimates
plot(em1,type="response",comparisons = T)
```



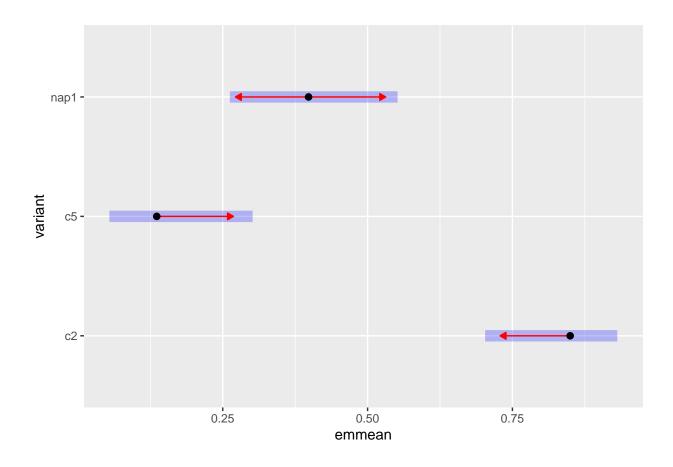
# Figure 3e

```
fig3e <- read.table("Fig3e",header = T)

m0 <- glm(data=fig3e,diffused~variant,family = quasibinomial)
anova(m0,test="LRT")

## Analysis of Deviance Table
##</pre>
```

```
## Model: quasibinomial, link: logit
##
## Response: diffused
##
## Terms added sequentially (first to last)
##
##
##
           Df Deviance Resid. Df Resid. Dev Pr(>Chi)
## NULL
                              59
                                     53.630
                                     31.413 7.703e-10 ***
## variant 2
               22.217
                              57
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
em1 <- emmeans(m0,pairwise~variant,type="response")</pre>
summary(em1)
## $emmeans
## variant prob
                     SE df asymp.LCL asymp.UCL
## c2
            0.850 0.0567 Inf
                                0.7029
                                           0.931
## c5
                                0.0543
                                           0.302
            0.136 0.0605 Inf
## nap1
           0.398 0.0759 Inf
                                0.2622
                                           0.552
##
## Confidence level used: 0.95
## Intervals are back-transformed from the logit scale
## $contrasts
## contrast odds.ratio
                             SE df null z.ratio p.value
## c2 / c5
                 35.879 24.392 Inf
                                      1
                                          5.266 <.0001
## c2 / nap1
                  8.543 4.662 Inf
                                           3.931 0.0002
                                       1
                  0.238 0.144 Inf
## c5 / nap1
                                       1 -2.374 0.0463
##
## P value adjustment: tukey method for comparing a family of 3 estimates
## Tests are performed on the log odds ratio scale
summary(em1)$contrasts$p.value
## [1] 4.174639e-07 2.491940e-04 4.625254e-02
plot(em1,type="response",comparisons = T)
```



## Figure 4a

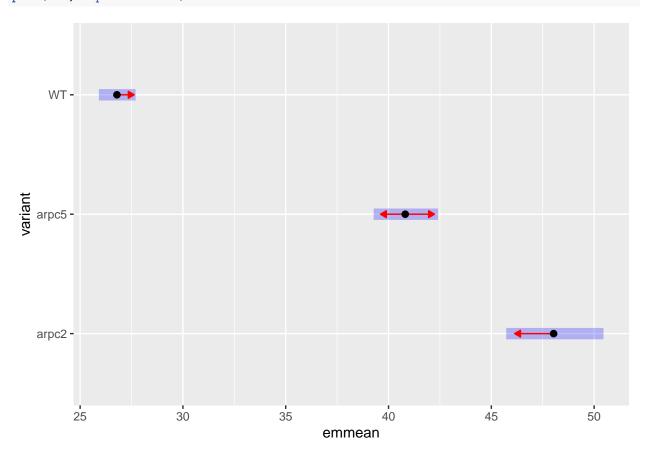
```
fig4a <- read.table("Fig4a",header = T)</pre>
m0 <- glm(data=fig4a,peroxisomes~variant,family = poisson)</pre>
anova(m0,test="LRT")
## Analysis of Deviance Table
##
## Model: poisson, link: log
##
## Response: peroxisomes
## Terms added sequentially (first to last)
##
##
           Df Deviance Resid. Df Resid. Dev Pr(>Chi)
##
                              225
## NULL
                                      1494.9
## variant 2
                 463.9
                              223
                                      1031.0 < 2.2e-16 ***
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
em1 <- emmeans(m0,pairwise~variant,type="response")</pre>
summary(em1)
```

## \$emmeans

```
SE df asymp.LCL asymp.UCL
##
   variant rate
##
    arpc2
           48.0 1.206 Inf
                                45.7
                                          50.5
                                39.3
                                          42.4
##
    arpc5
            40.8 0.799 Inf
            26.8 0.456 Inf
                                25.9
                                          27.7
##
## Confidence level used: 0.95
## Intervals are back-transformed from the log scale
##
## $contrasts
##
  contrast
                            SE df null z.ratio p.value
                 ratio
## arpc2 / arpc5 1.18 0.0375 Inf
                                     1
                                          5.114 <.0001
                                      1 19.244 <.0001
## arpc2 / WT
                   1.79 0.0544 Inf
## arpc5 / WT
                   1.52 0.0395 Inf
                                     1 16.235 <.0001
##
\#\# P value adjustment: tukey method for comparing a family of 3 estimates
## Tests are performed on the log scale
summary(em1)$contrasts$p.value
```

#### ## [1] 9.415278e-07 0.000000e+00 0.000000e+00

#### plot(em1,comparisons = T)



### Figure 4b

```
fig4b <- read.table("Fig4b",header = T)</pre>
m0 <- glm(data=fig4b, area~varianta,family = Gamma)</pre>
anova(m0,test="LRT")
## Analysis of Deviance Table
##
## Model: Gamma, link: inverse
## Response: area
##
## Terms added sequentially (first to last)
##
##
##
            Df Deviance Resid. Df Resid. Dev Pr(>Chi)
                              667
                                      251.32
## varianta 1 51.052
                                      200.26 < 2.2e-16 ***
                              666
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
m1 <- lm(data=fig4b, log(area)~varianta)
anova(m1)
## Analysis of Variance Table
## Response: log(area)
              Df Sum Sq Mean Sq F value
              1 46.161 46.161 154.21 < 2.2e-16 ***
## varianta
## Residuals 666 199.355
                           0.299
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
Figure 4c
fig4c <- read.table("Fig4c",header = T)</pre>
m0 <- glm(data=fig4c, area~variant,family = Gamma)</pre>
anova(m0,test="LRT")
## Analysis of Deviance Table
## Model: Gamma, link: inverse
##
## Response: area
## Terms added sequentially (first to last)
##
##
           Df Deviance Resid. Df Resid. Dev Pr(>Chi)
##
## NULL
                             967
                                     268.36
                             966
                                     227.02 < 2.2e-16 ***
## variant 1 41.335
## ---
```

```
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
m1 <- lm(data=fig4c, log(area)~variant)</pre>
anova(m1)
## Analysis of Variance Table
## Response: log(area)
             Df Sum Sq Mean Sq F value
             1 37.382 37.382 155.89 < 2.2e-16 ***
## Residuals 966 231.646
                         0.240
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
Figure 4d
fig4d <- read.table("Fig4d",header = T)
m1 <- lm(data=fig4d, log(length) ~ variant * treatment )</pre>
anova(m1)
## Analysis of Variance Table
## Response: log(length)
                    Df Sum Sq Mean Sq F value
                                                   Pr(>F)
                     3 4.3999 1.4666 40.5654 3.066e-16 ***
## variant
                     1 17.1267 17.1267 473.6989 < 2.2e-16 ***
## treatment
## variant:treatment 3 0.3684 0.1228
                                         3.3963
                                                 0.02161 *
                    83 3.0009 0.0362
## Residuals
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
em1 <- emmeans(m1, pairwise ~ treatment | variant, type="response")
summary(em1)
## $emmeans
## variant = arp2:
## treatment response SE df lower.CL upper.CL
## IBA
                 12.8 0.772 83
                                 11.40
                 30.4 1.828 83
                                  26.98
## mock
                                            34.3
##
## variant = arpc4:
## treatment response
                         SE df lower.CL upper.CL
## IBA
                 13.8 0.877 83
                                12.20
                 26.7 1.603 83
## mock
                                  23.66
                                            30.0
##
## variant = arpc5:
## treatment response
                         SE df lower.CL upper.CL
                 11.2 0.756 83
                                   9.83
## IBA
                                            12.8
## mock
                 26.1 1.655 83
                                  23.02
                                            29.6
##
## variant = WT:
## treatment response
                         SE df lower.CL upper.CL
## IBA
                 7.5 0.336 83
                                6.86
                                            8.2
## mock
                 20.4 0.940 83
                                  18.59
                                            22.3
```

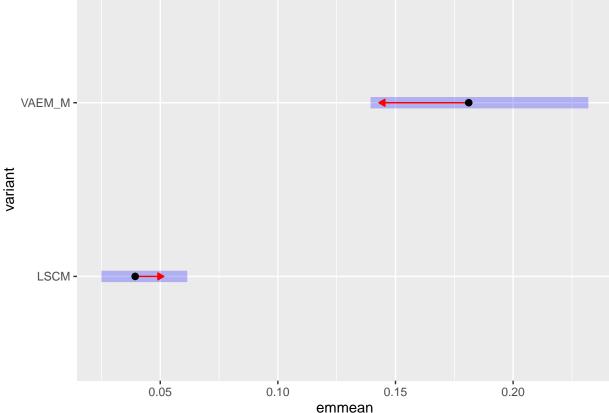
```
##
## Confidence level used: 0.95
## Intervals are back-transformed from the log scale
##
## $contrasts
## variant = arp2:
## contrast ratio
                        SE df null t.ratio p.value
## IBA / mock 0.422 0.0359 83
                                 1 -10.133 <.0001
##
## variant = arpc4:
## contrast ratio
                        SE df null t.ratio p.value
## IBA / mock 0.519 0.0454 83
                                 1 -7.503 <.0001
##
## variant = arpc5:
## contrast ratio
                        SE df null t.ratio p.value
## IBA / mock 0.430 0.0398 83
                                 1 -9.125 <.0001
##
## variant = WT:
## contrast ratio
                        SE df null t.ratio p.value
## IBA / mock 0.368 0.0237 83
                                 1 -15.545 <.0001
## Tests are performed on the log scale
summary(em1)$contrasts$p.value
## [1] 3.575741e-16 6.307242e-11 3.677664e-14 2.636529e-26
Figure 4h
fig4h <- read.table("Fig4h",header = T)</pre>
m0 <- glm(data=fig4h,coloc~variant,family = quasibinomial)</pre>
anova(m0,test="LRT")
## Analysis of Deviance Table
## Model: quasibinomial, link: logit
## Response: coloc
## Terms added sequentially (first to last)
##
##
##
          Df Deviance Resid. Df Resid. Dev Pr(>Chi)
## NULL
                            247
                                    75.953
## variant 1
                            246
                                    62.509 3.401e-10 ***
             13.444
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

```
## $emmeans
## variant prob SE df asymp.LCL asymp.UCL
## LSCM 0.0393 0.00906 Inf 0.025 0.0615
```

em1 <- emmeans(m0,pairwise~variant,type="response")</pre>

summary(em1)

```
## VAEM_M 0.1811 0.02357 Inf
                                   0.139
                                            0.2320
##
## Confidence level used: 0.95
## Intervals are back-transformed from the logit scale
## $contrasts
## contrast
                  odds.ratio
                                 SE df null z.ratio p.value
                                           1 -5.864 <.0001
## LSCM / VAEM_M
                     0.185 0.0532 Inf
##
\mbox{\tt \#\#} Tests are performed on the log odds ratio scale
summary(em1)$contrasts$p.value
## [1] 4.51307e-09
plot(em1,type="response",comparisons = T)
```



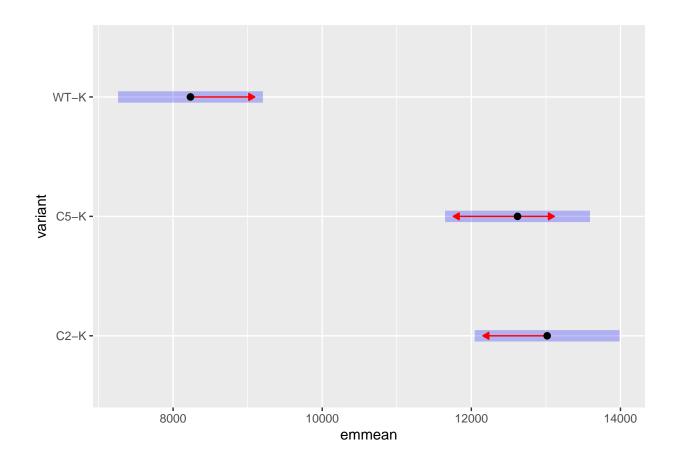
# Figure 4o

```
fig4o <- read.table("Fig4o",header = T)
anova(lm(data=fig4o,tm~autophagosome))
## Analysis of Variance Table
##
## Response: tm</pre>
```

```
## Df Sum Sq Mean Sq F value Pr(>F)
## autophagosome 1 727685 727685 15.847 0.000299 ***
## Residuals 38 1744983 45921
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

## Figure S4a

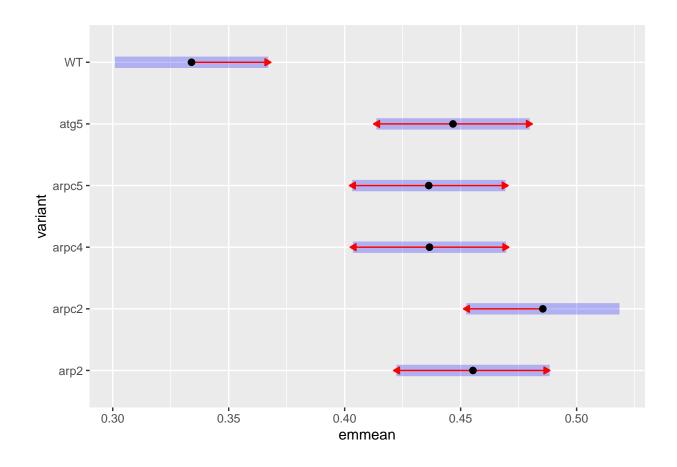
```
figS4a <- read.table("FigS4a",header = T)</pre>
m1 <- aov(data=figS4a,density~variant)</pre>
summary(m1)
##
                   Sum Sq Mean Sq F value
                                             Pr(>F)
               2 42287384 21143692
                                    44.71 0.000249 ***
## variant
## Residuals
               6 2837368
                            472895
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
em1 <- emmeans(m1,pairwise~variant,type="response")</pre>
summary(em1)
## $emmeans
## variant emmean SE df lower.CL upper.CL
           13017 397 6
                            12046
## C5-K
            12619 397 6
                             11648
                                     13591
## WT-K
             8233 397 6
                             7261
                                      9204
##
## Confidence level used: 0.95
##
## $contrasts
## contrast
                    estimate SE df t.ratio p.value
## (C2-K) - (C5-K)
                        398 561 6 0.708 0.7678
                                   8.521 0.0004
## (C2-K) - (WT-K)
                        4784 561 6
                       4387 561 6 7.812 0.0006
## (C5-K) - (WT-K)
## P value adjustment: tukey method for comparing a family of 3 estimates
summary(em1)$contrasts$p.value
## [1] 0.7678493786 0.0003515171 0.0005682736
plot(em1,type="response",comparisons = T)
```



## Figure S4c

```
figS4c <- read.table("FigS4c",header = T)</pre>
m0 <- aov(data=figS4c,fluorescence~variant)</pre>
summary(m0)
##
                Df Sum Sq Mean Sq F value
                                             Pr(>F)
                 5 0.2387 0.04775
                                      9.56 1.67e-07 ***
## variant
               102 0.5095 0.00499
## Residuals
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
em1 <- emmeans(m0,pairwise~variant,type="response")</pre>
summary(em1)
## $emmeans
## variant emmean
                       SE df lower.CL upper.CL
## arp2
             0.455 0.0167 102
                                  0.422
                                           0.488
## arpc2
             0.485 0.0167 102
                                  0.452
                                           0.518
## arpc4
             0.437 0.0167 102
                                  0.404
                                           0.470
## arpc5
             0.436 0.0167 102
                                  0.403
                                           0.469
             0.447 0.0167 102
##
   atg5
                                  0.414
                                           0.480
## WT
             0.334 0.0167 102
                                  0.301
                                           0.367
##
## Confidence level used: 0.95
```

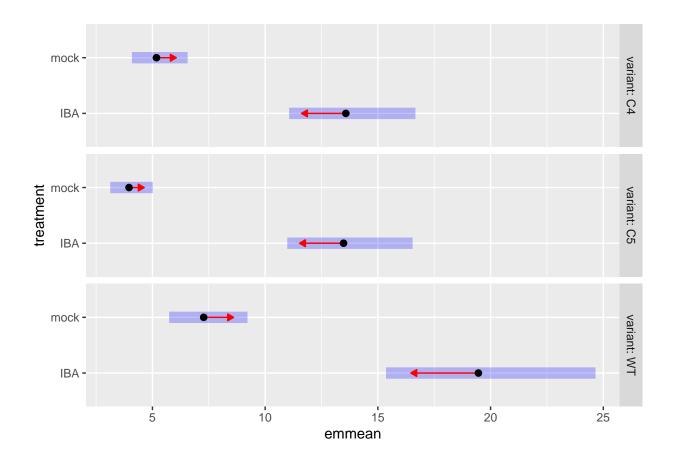
```
##
## $contrasts
## contrast
                              SE df t.ratio p.value
                 estimate
## arp2 - arpc2 -0.030120 0.0236 102 -1.279 0.7960
                0.018739 0.0236 102
##
   arp2 - arpc4
                                     0.795 0.9677
## arp2 - arpc5 0.019057 0.0236 102 0.809 0.9653
## arp2 - atg5
                 0.008612 0.0236 102 0.366 0.9991
## arp2 - WT
                 0.121348 0.0236 102 5.151 <.0001
##
   arpc2 - arpc4 0.048859 0.0236 102 2.074 0.3091
## arpc2 - arpc5 0.049176 0.0236 102 2.087 0.3021
## arpc2 - atg5  0.038732  0.0236  102  1.644  0.5716
                 0.151468 0.0236 102 6.430 <.0001
## arpc2 - WT
   arpc4 - arpc5 0.000318 0.0236 102 0.013 1.0000
##
## arpc4 - atg5 -0.010127 0.0236 102 -0.430 0.9981
## arpc4 - WT
                 0.102609 0.0236 102
                                     4.356 0.0004
##
   arpc5 - atg5 -0.010445 0.0236 102 -0.443 0.9978
##
                 0.102292 0.0236 102 4.342 0.0005
   arpc5 - WT
                  0.112736 0.0236 102 4.786 0.0001
## atg5 - WT
##
## P value adjustment: tukey method for comparing a family of 6 estimates
summary(em1)$contrasts$p.value
## [1] 7.960011e-01 9.677057e-01 9.652896e-01 9.991227e-01 1.850841e-05
  [6] 3.091474e-01 3.020741e-01 5.715834e-01 6.219121e-08 1.000000e+00
## [11] 9.980844e-01 4.449431e-04 9.977795e-01 4.681686e-04 8.319533e-05
plot(em1,type="response",comparisons = T)
```



## Figure S4d

```
figS4d <- read.table("FigS4d",header = T)</pre>
m1 <- lm(data=figS4d, log(primcm)~variant*treatment)</pre>
anova(m1)
## Analysis of Variance Table
##
## Response: log(primcm)
                     Df Sum Sq Mean Sq F value
                                                    Pr(>F)
## variant
                      2 0.5365 0.2682
                                        7.3526 0.006564 **
## treatment
                      1 5.5441 5.5441 151.9640 6.616e-09 ***
## variant:treatment 2 0.0708 0.0354
                                        0.9699 0.403183
## Residuals
                     14 0.5108 0.0365
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
m2 <- lm(data=figS4d, log(primcm)~variant+treatment)</pre>
anova(m2)
## Analysis of Variance Table
## Response: log(primcm)
             Df Sum Sq Mean Sq F value
                                            Pr(>F)
```

```
2 0.5365 0.2682 7.3803 0.005358 **
## treatment 1 5.5441 5.5441 152.5372 1.354e-09 ***
## Residuals 16 0.5815 0.0363
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
em1 <- emmeans(m1, pairwise ~ treatment | variant, type="response")</pre>
summary(em1)
## $emmeans
## variant = C4:
## treatment response
                        SE df lower.CL upper.CL
              13.58 1.297 14
                               11.06
                                          16.67
## mock
                5.18 0.571 14
                                  4.09
                                           6.56
##
## variant = C5:
## treatment response SE df lower.CL upper.CL
## IBA
               13.48 1.287 14
                                10.98
                                         16.54
## mock
                 3.96 0.437 14
                                  3.13
                                           5.02
##
## variant = WT:
## treatment response SE df lower.CL upper.CL
                19.46 2.146 14
                                 15.36
                                          24.65
## mock
                7.27 0.802 14
                                 5.74
                                           9.22
## Confidence level used: 0.95
## Intervals are back-transformed from the log scale
##
## $contrasts
## variant = C4:
## contrast ratio
                      SE df null t.ratio p.value
## IBA / mock 2.62 0.382 14 1 6.607 <.0001
##
## variant = C5:
## contrast ratio
                       SE df null t.ratio p.value
## IBA / mock 3.40 0.496 14
                              1 8.392 <.0001
##
## variant = WT:
## contrast ratio
                       SE df null t.ratio p.value
## IBA / mock 2.68 0.417 14
                               1 6.310 <.0001
## Tests are performed on the log scale
summary(em1)$contrasts$p.value
## [1] 1.175403e-05 7.807748e-07 1.923172e-05
plot(em1, comparison=T)
```



### Figure S4e

```
figS4e <- read.table("FigS4e",header = T)

m1 <- betareg(data=figS4e,I(percarea/100)~variant)
lrtest(m1)

## Likelihood ratio test

##

## Model 1: I(percarea/100) ~ variant

## Model 2: I(percarea/100) ~ 1

## #Df LogLik Df Chisq Pr(>Chisq)

## 1 5 83.760

## 2 2 82.934 -3 1.6514 0.6478
```

# Figure 5a-d

```
fig5ad <- read.table("Fig5ad",header = T)

# Counts were normalized by scanning areas (~ divided approx. by 2), thus the
# multiplication by 2 was necessary to prevent the loss of complexity in data
# after rounding (to get discrete values for Poisson model).
m1 <- glm(data=fig5ad,round(2*autophagosomes)~variant,family = poisson)
anova(m1,test="LRT")</pre>
```

```
## Analysis of Deviance Table
##
## Model: poisson, link: log
## Response: round(2 * autophagosomes)
##
## Terms added sequentially (first to last)
##
##
##
           Df Deviance Resid. Df Resid. Dev Pr(>Chi)
## NULL
                             242
                                     797.00
                61.964
                                     735.04 3.499e-15 ***
## variant 1
                             241
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
m1 <- glm(data=fig5ad,round(2*arposomes)~variant,family = poisson)
anova(m1,test="LRT")
## Analysis of Deviance Table
## Model: poisson, link: log
##
## Response: round(2 * arposomes)
##
## Terms added sequentially (first to last)
##
           Df Deviance Resid. Df Resid. Dev Pr(>Chi)
##
## NULL
                             242
                                     1252.3
## variant 1
                192.24
                             241
                                     1060.0 < 2.2e-16 ***
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
m1 <- glm(data=fig5ad,round(2*colocalization)~variant,family = poisson)</pre>
anova(m1,test="LRT")
## Analysis of Deviance Table
##
## Model: poisson, link: log
## Response: round(2 * colocalization)
## Terms added sequentially (first to last)
##
##
##
           Df Deviance Resid. Df Resid. Dev Pr(>Chi)
## NULL
                             242
                                     254.04
                             241
                                     244.92 0.002533 **
## variant 1
              9.1169
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
m1 <- glm(data=fig5ad,col_aut~variant,family = quasibinomial)</pre>
anova(m1,test="LRT")
## Analysis of Deviance Table
```

```
## Model: quasibinomial, link: logit
##
## Response: col_aut
##
## Terms added sequentially (first to last)
##
##
##
Df Deviance Resid. Df Resid. Dev Pr(>Chi)
## NULL 242 88.322
## variant 1 0.033692 241 88.288 0.8026
```

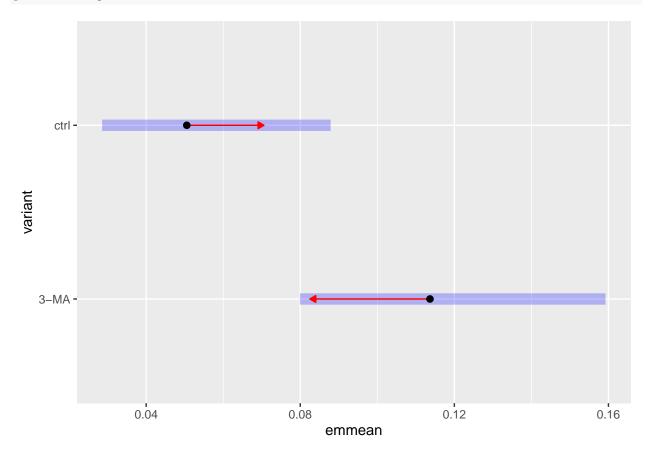
### Figure 5e-h

```
fig5eh <- read.table("Fig5efh",header = T)</pre>
m1 <- glm(data=fig5eh,round(autophagosomes)~variant,family = poisson)</pre>
anova(m1,test="LRT")
## Analysis of Deviance Table
## Model: poisson, link: log
##
## Response: round(autophagosomes)
## Terms added sequentially (first to last)
##
##
##
           Df Deviance Resid. Df Resid. Dev Pr(>Chi)
                              60
                                     250.71
## NULL
                31.797
                              59
                                     218.91 1.711e-08 ***
## variant 1
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
m1 <- glm(data=fig5eh,round(arposomes)~variant,family = poisson)
anova(m1,test="LRT")
## Analysis of Deviance Table
## Model: poisson, link: log
##
## Response: round(arposomes)
## Terms added sequentially (first to last)
##
##
           Df Deviance Resid. Df Resid. Dev Pr(>Chi)
##
## NULL
                              60
                                    1054.90
## variant 1
              377.33
                              59
                                     677.58 < 2.2e-16 ***
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
m1 <- glm(data=fig5eh,round(colocalization)~variant,family = poisson)
anova(m1,test="LRT")
```

```
## Model: poisson, link: log
## Response: round(colocalization)
## Terms added sequentially (first to last)
##
           Df Deviance Resid. Df Resid. Dev Pr(>Chi)
                              60
## NULL
                                     96.910
## variant 1
                20.535
                              59
                                     76.375 5.855e-06 ***
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
m1 <- glm(data=fig5eh,col_aut~variant,family = quasibinomial)</pre>
anova(m1,test="LRT")
## Analysis of Deviance Table
## Model: quasibinomial, link: logit
## Response: col_aut
## Terms added sequentially (first to last)
##
##
           Df Deviance Resid. Df Resid. Dev Pr(>Chi)
                              60
## NULL
                                     8.5090
## variant 1 0.81906
                              59
                                     7.6899 0.01129 *
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
em1 <- emmeans(m1,pairwise~variant,type="response")</pre>
summary(em1)
## $emmeans
## variant
              prob
                       SE df asymp.LCL asymp.UCL
            0.1137 0.0200 Inf
                                 0.0799
## 3-MA
                                           0.1592
                                 0.0286
            0.0505 0.0145 Inf
                                           0.0879
##
## Confidence level used: 0.95
## Intervals are back-transformed from the logit scale
## $contrasts
   contrast
                  odds.ratio
                                SE df null z.ratio p.value
## (3-MA) / ctrl
                        2.41 0.873 Inf
                                              2.427 0.0152
                                          1
## Tests are performed on the log odds ratio scale
summary(em1)$contrasts$p.value
```

## [1] 0.01521345





# Figure 5i-k

```
fig5ijk <- read.table("Fig5ijk",header = T)</pre>
m1 <- glm(data=fig5ijk,arposomes~variant,family = poisson)</pre>
anova(m1,test="LRT")
## Analysis of Deviance Table
## Model: poisson, link: log
##
## Response: arposomes
## Terms added sequentially (first to last)
##
##
           Df Deviance Resid. Df Resid. Dev Pr(>Chi)
##
## NULL
                              24
                                    2543.39
## variant 1
              1995.2
                              23
                                     548.14 < 2.2e-16 ***
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
```

```
m1 <- glm(data=fig5ijk,peroxisomes~variant,family = poisson)</pre>
anova(m1,test="LRT")
## Analysis of Deviance Table
## Model: poisson, link: log
## Response: peroxisomes
## Terms added sequentially (first to last)
##
##
           Df Deviance Resid. Df Resid. Dev Pr(>Chi)
## NULL
                              24
                                     3651.3
                2882.5
                              23
                                      768.8 < 2.2e-16 ***
## variant 1
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
m1 <- betareg(data=fig5ijk,I(ap_ratio/100)~variant)</pre>
lrtest(m1)
## Likelihood ratio test
## Model 1: I(ap_ratio/100) ~ variant
## Model 2: I(ap_ratio/100) ~ 1
## #Df LogLik Df Chisq Pr(>Chisq)
## 1
       3 27.515
       2 14.816 -1 25.398 4.665e-07 ***
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
Figure S5
figS5 <- read.table("FigS5",header = T)</pre>
m1 <- glm(data=figS5,coloc~variant,family = quasibinomial)</pre>
anova(m1,test="LRT")
## Analysis of Deviance Table
## Model: quasibinomial, link: logit
## Response: coloc
## Terms added sequentially (first to last)
##
##
##
           Df Deviance Resid. Df Resid. Dev Pr(>Chi)
## NULL
                             304
                                     106.58
## variant 2
              16.415
                             302
                                      90.17 2.698e-10 ***
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
```

```
em1 <- emmeans(m1,pairwise~variant,type="response")</pre>
summary(em1)
## $emmeans
  variant
                        SE df asymp.LCL asymp.UCL
            prob
## LSCM
            0.0393 0.00947 Inf
                                 0.0244
                                            0.0627
## VAEM_A 0.1741 0.03066 Inf
                                 0.1219
                                            0.2425
##
  VAEM_M 0.1811 0.02464 Inf
                                 0.1377
                                            0.2345
##
## Confidence level used: 0.95
## Intervals are back-transformed from the logit scale
##
## $contrasts
                    odds.ratio
## contrast
                                  SE df null z.ratio p.value
## LSCM / VAEM_A
                        0.194 0.0639 Inf
                                             1 -4.981 <.0001
## LSCM / VAEM_M
                        0.185 0.0557 Inf
                                               -5.610 <.0001
##
  VAEM_A / VAEM_M
                        0.953 0.2576 Inf
                                             1 -0.178 0.9827
## P value adjustment: tukey method for comparing a family of 3 estimates
## Tests are performed on the log odds ratio scale
summary(em1)$contrasts$p.value
```



