## 2025079\_dharma\_plots

#### 2025-07-09

### Using DHARMa package to evaluate the fit of our negative binomial model

- Model format for all data from all journals (N = 47353)
  - MASS::glm.nb(is.referenced.by.count~ da\_factor + log(age.in.months) + container.title + container.title\*da\_factor + log(age.in.months)\*da\_factor + container.title\*log(age.in.months) + log(age.in.months)\*da\_factor\*container.title, data = nsd\_yes\_metadata, link = log)
  - Data nsd\_yes\_metadata was filtered to remove all NAs from variables da\_factor, age.in.months, and container.title to allow for some of the below visualizations.
  - See below for number of papers from each journal.

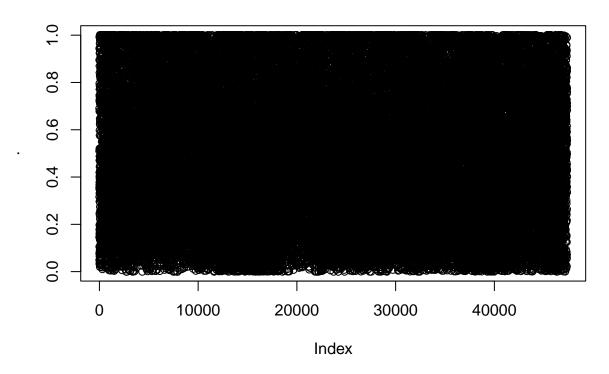
n
8613
6578
5691
4656
4577
4369
3223
2900
2438
1854
1406
1041
7

#### Plot Residuals

- Graph is very busy with all 47K observations
- see 2nd plot for 5K observations for a smaller group to examine (index 25000-30000)
- Residuals look like an amorphous blob as suggested by Abner@CSCAR

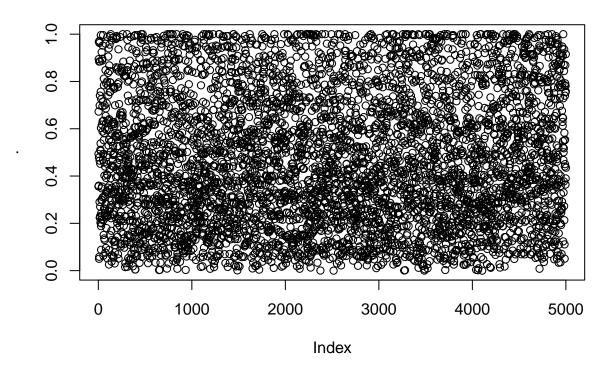
```
simulationOutput <- simulateResiduals(fittedModel = total_model, plot = F)
residuals(simulationOutput) %>%
  plot(main = "Residuals plotted by numerical index")
```

# Residuals plotted by numerical index



```
residuals(simulationOutput)[25000:30000] %>%
plot(main = "Residuals for observation indices 25000 - 3000")
```

## Residuals for observation indices 25000 - 3000



### Plots for Residuals - QQ and DHARMa

#### QQ Plot

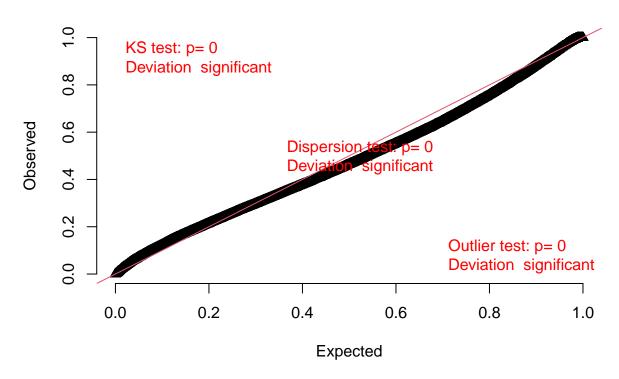
- KS Test = Two sample Kolmogorov-Smirnov (KS) Test
  - This function tests the overall uniformity of the simulated residuals in a DHARMa object Deviation is significant between the expected residuals and the actual observed residuals.
  - "If the P value is small, conclude that the two groups were sampled from populations with different distributions." -Prism help page
- Dispersion Test
  - This function performs simulation-based tests for over/underdispersion
  - Over / underdispersion means that the observed data is more / less dispersed than expected under the fitted model.
  - Deviation is significant between the observed data and fitted model.
- Outlier Test
  - This function tests if the number of observations outside the simulation envelope are larger or smaller than expected
  - Methods generate a null expectation, and then test for an excess or lack of outliers. Per default, testOutliers() looks for both, so if you get a significant p-value, you have to check if you have to many or too few outliers.

- See Outlier test for distribution of outliers. - Many at 1.0 residual.

#### plotQQunif(simulationOutput)

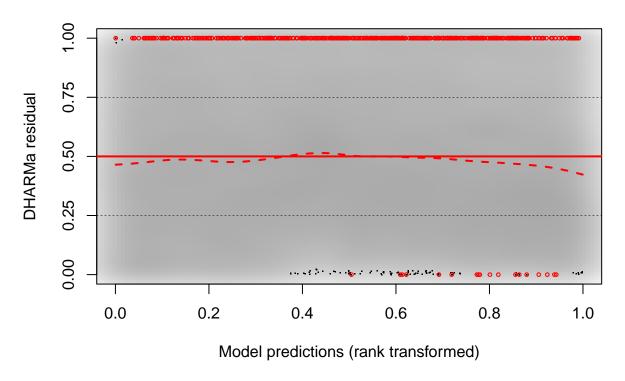
## DHARMa:testOutliers with type = binomial may have inflated Type I error rates for integer-valued dis

## QQ plot residuals



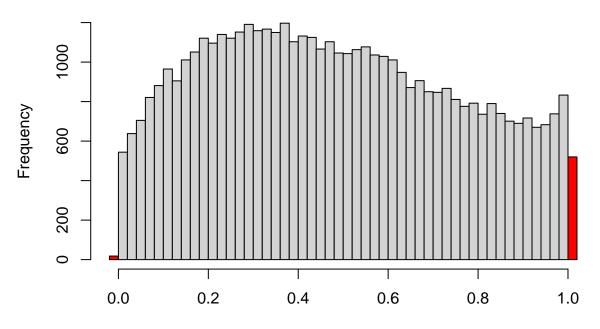
plotResiduals(simulationOutput)

# DHARMa residual vs. predicted



testOutliers(simulationOutput, type = "bootstrap")

## **Outlier test significant**



Residuals (outliers are marked red)

```
##
## DHARMa bootstrapped outlier test
##
## data: simulationOutput
## outliers at both margin(s) = 538, observations = 47353, p-value <
## 2.2e-16
## alternative hypothesis: two.sided
## percent confidence interval:
## 0.006957848 0.008756045
## sample estimates:
## outlier frequency (expected: 0.00784047473232953 )
## 0.01136148</pre>
```