

Brunswick Point Peep Models

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Roberts Bank, Delta, BC, is situated in the great Pacific Flyway and serves as an important stopover for peeps migrating north in the spring. It therefore hosts a large seasonal population of peeps: namely, Western sandpiper (WESA) and Dunlin (DUNL), that rely on the seasonal nutritional bounty provided by the Fraser River delta.

This document describes the suite of models used to estimate yearly changes in spatial distribution and abundance of peeps in the Roberts Bank estuary.

There are two datasets used in this modelling pipeline:

1. **sr**, which contains species ratio data (WESA:DUNL)
2. **dat**, which contains bird counts + environmental covariates

For an interactive interface with the **dat** dataset used in this document, see <https://popovs.shinyapps.io/peepr/>.

1 Binomial species ratio model

The daily ratio of Western sandpiper (WESA) to Dunlin (DUNL) across the entire study period is first modelled using a dataset of known species ratios (species ratios are not measured during every survey).

The ratios are modelled using a binomial generalized linear mixed model (binomial GLMM). The resulting predicted ratios are then used to estimate the number of WESA vs. DUNL per day.

1.1 sr data summary

```
##   survey_date           wesa          dunl          total
## Min.   :1997-04-21   Min.   : 0.0   Min.   : 0.0   Min.   : 48.0
## 1st Qu.:2005-04-20   1st Qu.:126.8   1st Qu.: 42.0   1st Qu.:287.2
## Median :2009-04-23   Median :355.0   Median : 93.0   Median :524.5
## Mean   :2009-09-05   Mean   :470.1   Mean   :176.7   Mean   :646.8
## 3rd Qu.:2015-04-16   3rd Qu.:719.0   3rd Qu.:203.5   3rd Qu.:943.8
## Max.   :2022-05-04   Max.   :2605.0  Max.   :5000.0  Max.   :5047.0
##
##      p_wesa        p_dunl       year    julian_day
## Min.   : 0.00   Min.   : 0.00   2006   : 23   Min.   :101.0
## 1st Qu.: 53.04   1st Qu.: 8.83   2019   : 23   1st Qu.:112.0
## Median : 81.32   Median :18.68   2005   : 18   Median :117.0
## Mean   : 68.94   Mean   :31.06   2007   : 18   Mean   :117.6
## 3rd Qu.: 91.17   3rd Qu.:46.96   2008   : 18   3rd Qu.:123.0
## Max.   :100.00   Max.   :100.00   2012   : 18   Max.   :137.0
##                                     (Other):200
##
##      dos.V1
## Min.   :-2.2763470
## 1st Qu.:-0.7798519
## Median : -0.0996268
## Mean   : -0.0221924
## 3rd Qu.: 0.7166433
## Max.   : 2.6212735
##
```

1.2 Five models are built and compared

Response variable:

- y - WESA:DUNL ratio

Predictor variables:

- dos - day of season (recentered/scaled Julian date)
- year - year of survey

```
## Data: sr
## Models:
## lme5: y ~ 1 + (1 | year)
## lme4: y ~ dos + (1 | year)
## lme2: y ~ dos + I(dos^2) + (1 | year)
```

```

## lme3: y ~ dos + (dos | year)
## lme1: y ~ dos + I(dos^2) + (dos + I(dos^2) | year)
##      npar   AIC   BIC logLik deviance Chisq Df Pr(>Chisq)
## lme5    2 64756 64764 -32376     64752
## lme4    3 25804 25815 -12899    25798 38954.1  1     <2e-16 ***
## lme2    4 16295 16310  -8144    16287  9510.6  1     <2e-16 ***
## lme3    5 22521 22540 -11255    22511    0.0  1          1
## lme1    9 11083 11117  -5532    11065 11446.2  4     <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

The best-fit model is `lme1`. The residuals from `lme1` are appended to the `sr` dataset and another model is re-fit in order to estimate overdispersion. Because the standard deviation of the residuals is < 1 , the model is deemed an appropriate candidate for predicting daily species ratios.

```

## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: y ~ dos + I(dos^2) + (dos + I(dos^2) | year) + (1 | resids)
## Data: sr
##      AIC      BIC logLik deviance df.resid
## 3409.879 3447.499 -1694.939  3389.879     308
## Random effects:
## Groups Name        Std.Dev. Corr
## resids (Intercept) 0.94716
## year    (Intercept) 0.65139
## dos      dos        0.06454 -1.00
##           I(dos^2)  0.41479 -0.65  0.68
## Number of obs: 318, groups: resids, 318; year, 24
## Fixed Effects:
## (Intercept)      dos      I(dos^2)
##       1.8826     1.3770    -0.8085

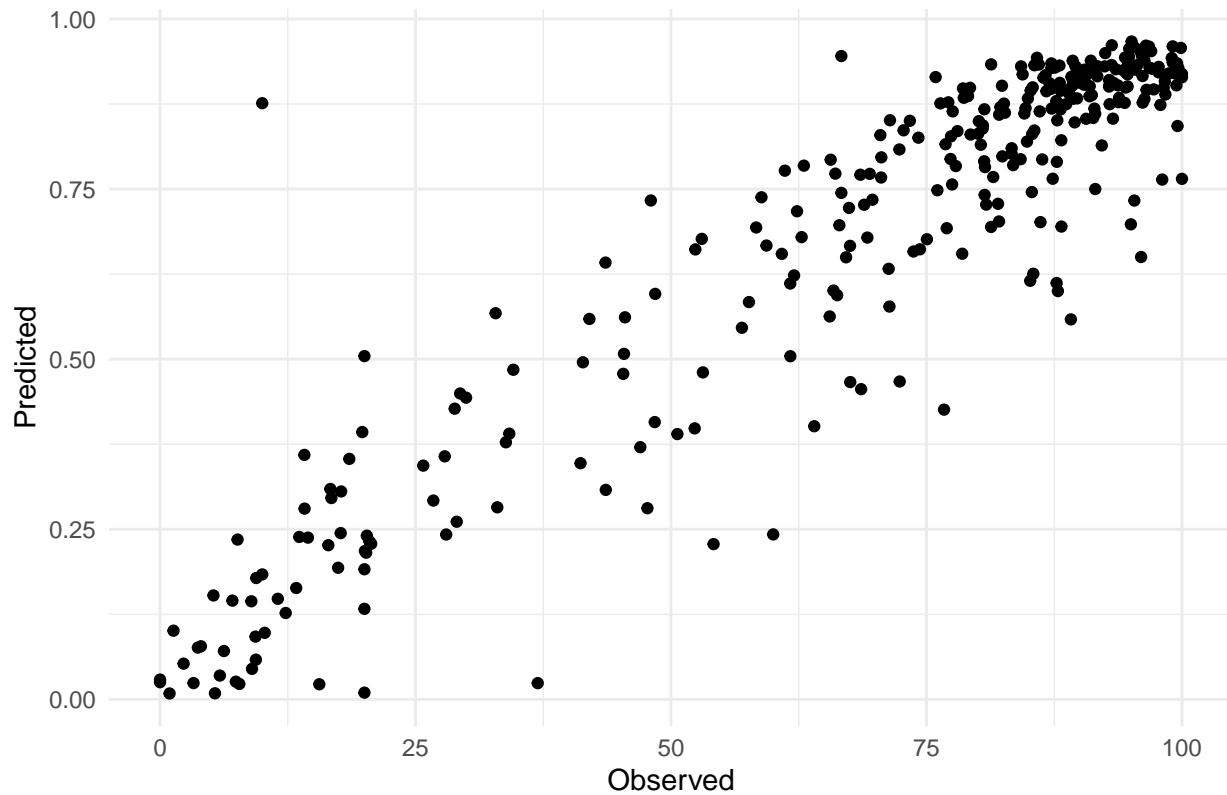
```

1.2.1 Summary of best fit model (lme1)

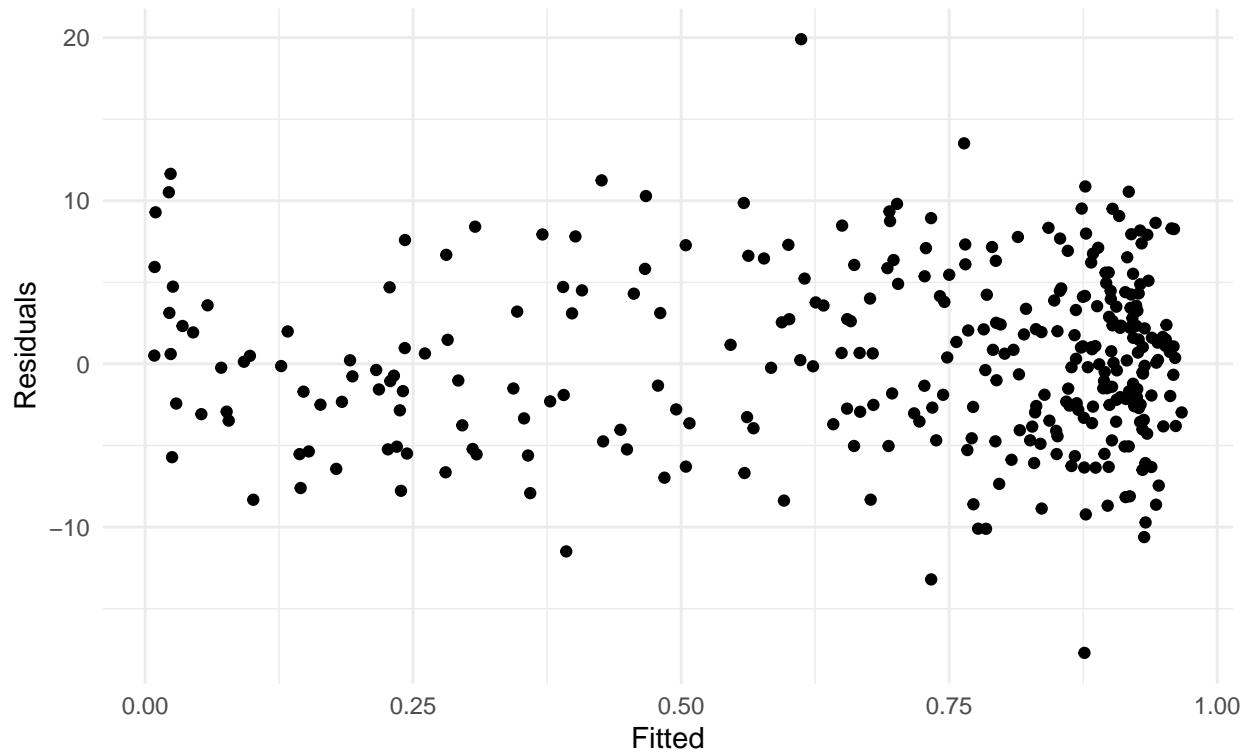
```
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: y ~ dos + I(dos^2) + (dos + I(dos^2) | year)
## Data: sr
##
##      AIC      BIC  logLik deviance df.resid
## 11082.7 11116.6 -5532.4 11064.7     309
##
## Scaled residuals:
##      Min      1Q  Median      3Q     Max
## -23.5674 -3.5412  0.2172  3.6901 21.7396
##
## Random effects:
## Groups Name        Variance Std.Dev. Corr
## year   (Intercept) 0.6045   0.7775
## dos       0.4629   0.6803   0.18
## I(dos^2)  0.6650   0.8155  -0.71 -0.53
## Number of obs: 318, groups: year, 24
##
## Fixed effects:
##             Estimate Std. Error z value Pr(>|z|)
## (Intercept) 1.7477    0.1593 10.974 < 2e-16 ***
## dos         1.3365    0.1398  9.558 < 2e-16 ***
## I(dos^2)   -0.8168    0.1676 -4.874 1.09e-06 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
## (Intr) dos
## dos     0.177
## I(dos^2) -0.705 -0.522
```

1.2.2 Check assumptions of best fit model

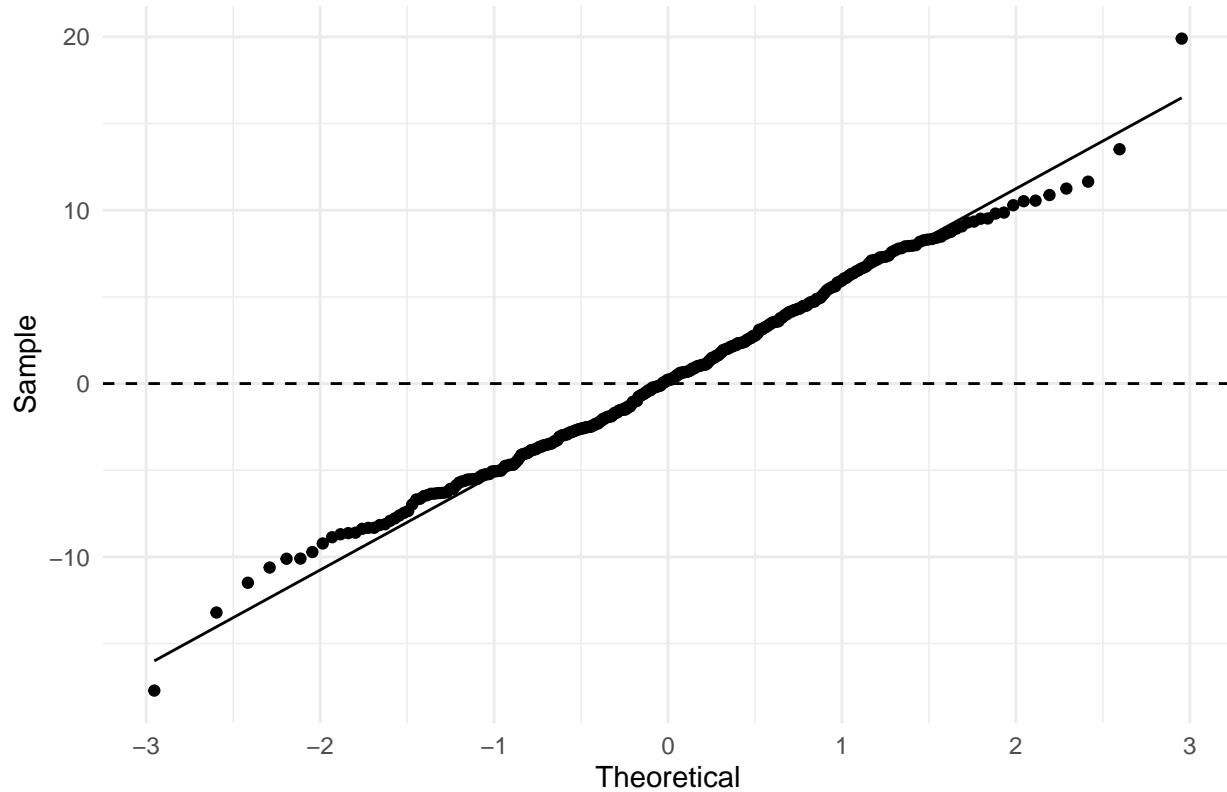
Observed vs. Predicted values

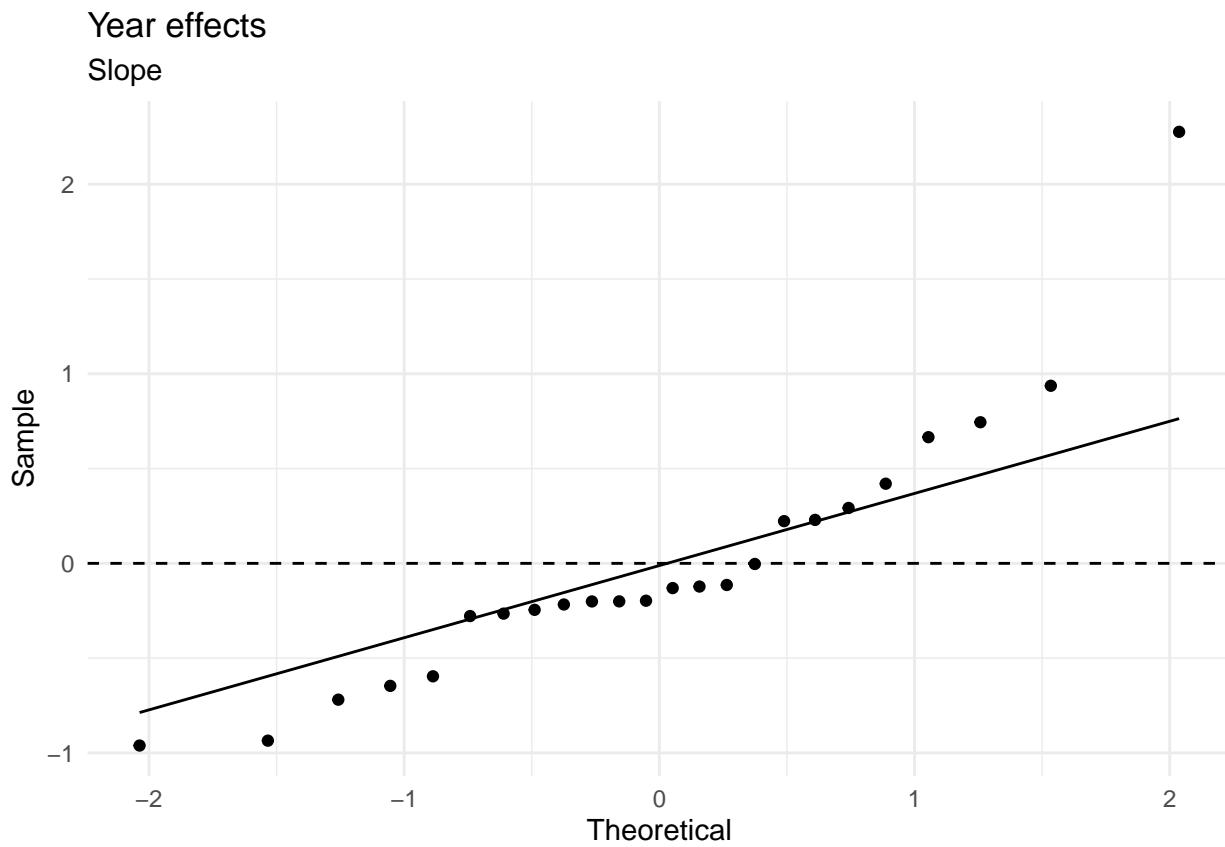
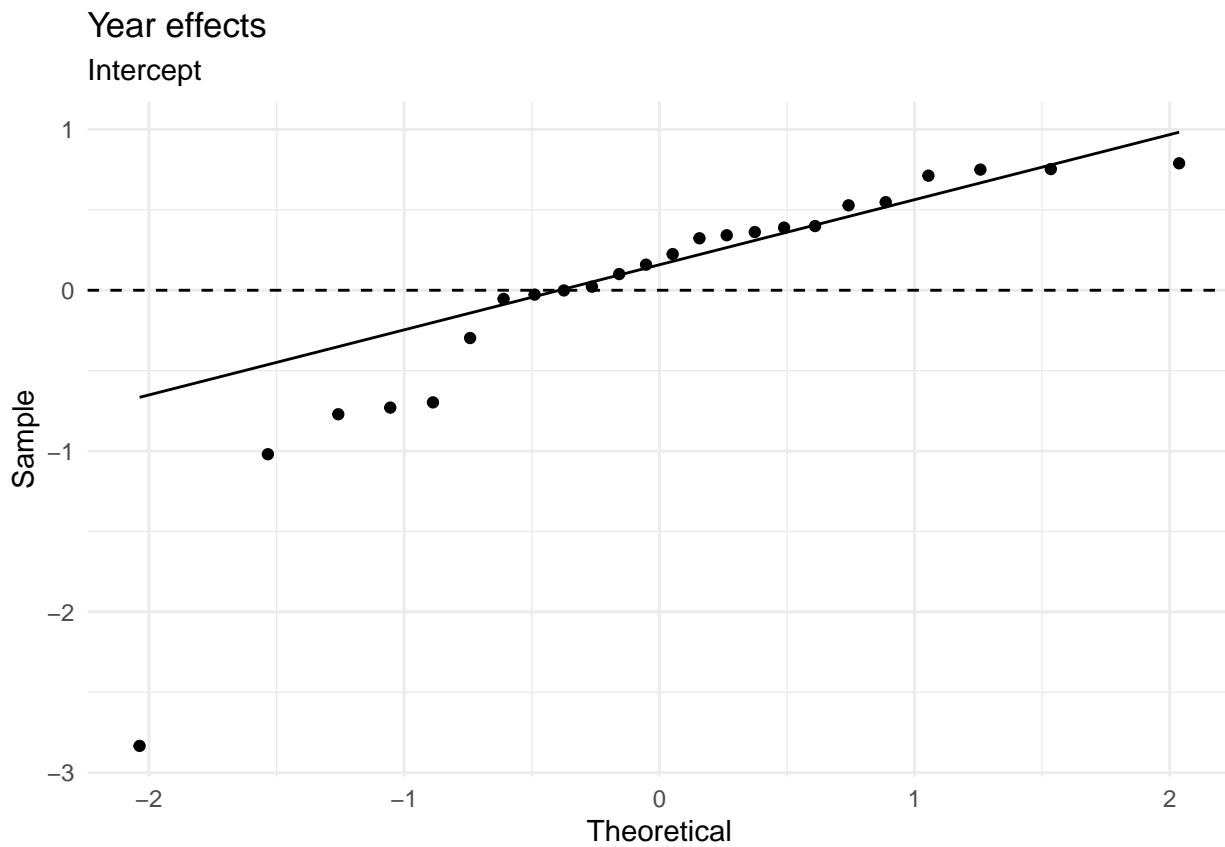


Heteroskedasticity Fitted values vs. Residuals



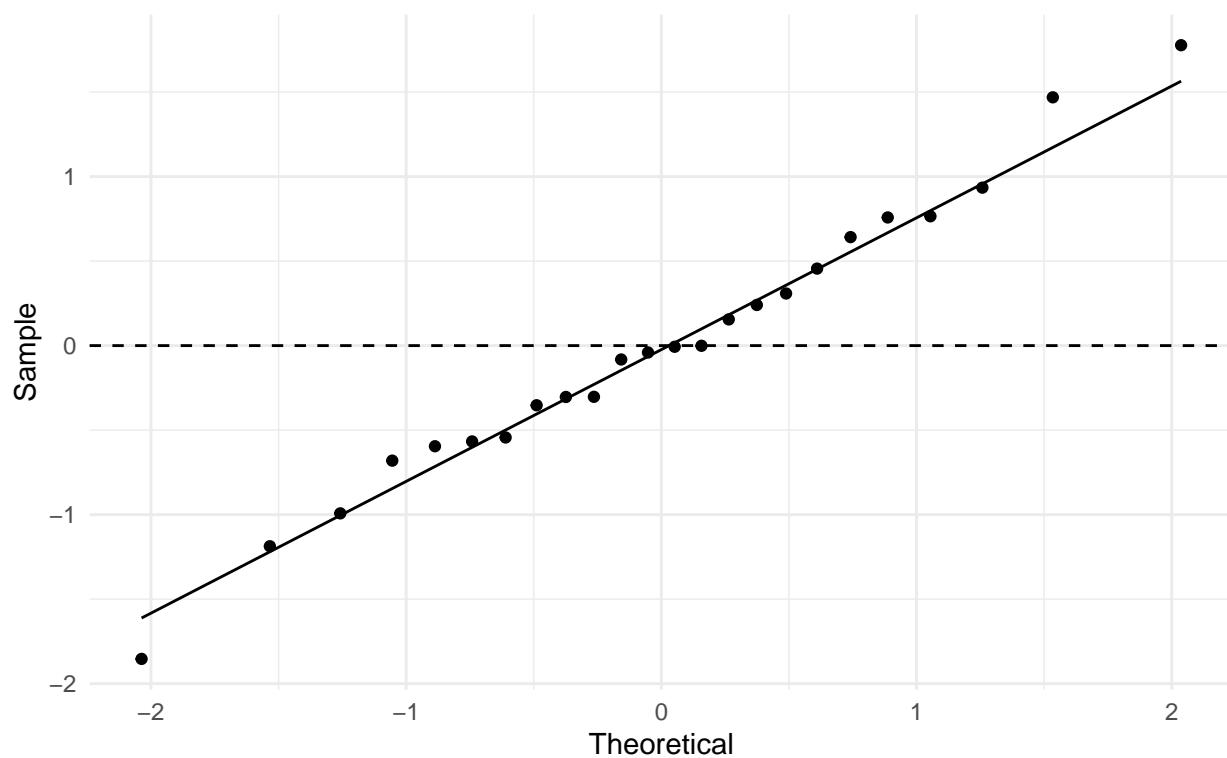
Quantile–Quantile





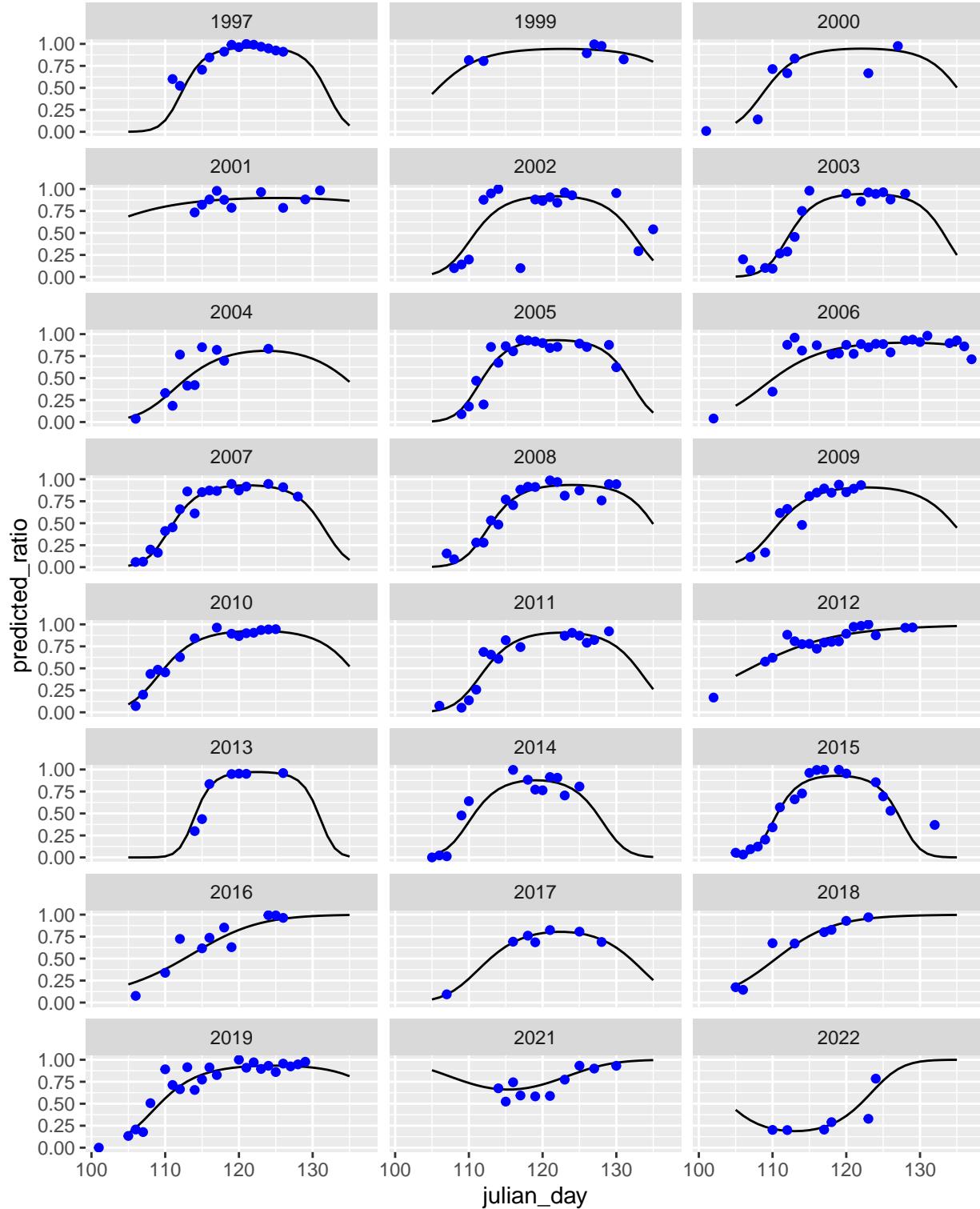
Year effects

Slope quadratic



1.3 Predict WESA/DUNL population

Using the derived binomial GLMM above, we will predict the amount of WESA and DUNL each day. For any years that are missing from the bGLMM we will assign the mean proportion of WESA:DUNL.



2 Model data summary

The primary goal of this analysis is to determine whether melt flow regimes alter the north-south distribution of peeps. Birds are surveyed from a route of standardized survey stations. Any birds counted in or north of the “view corner” are labelled as **N**, while any birds counted south of the “view corner” are labelled as **S** for this analysis.¹

The **dat** dataset, which contains survey data and associated environmental covariates, will be used in this model. The **dat** dataset can be explored in greater detail in the PeepR Shiny app.

¹For now, this includes birds counted in Canoe Pass, but those will be excluded in a later analysis to compare results.

2.0.1 dat filtering

The `dat` dataset underwent some filtering steps to exclude unwanted data prior to this analysis. First, only data that was originally included in Canham et al. (2021) is extracted from the `bppeeps` database. While survey data oftentimes includes multiple sweeps, low-quality or reconnaissance sweeps were excluded from the daily totals used in Canham et al. (2021). As such, they were excluded from this dataset as well.

The total bird count for several surveys were obtained by taking the average total count from multiple sweeps. In these cases, it is difficult to obtain accurate location-level numbers because birds often moved locations between sweeps. These records are excluded for now (45 survey dates out of a total of 538).

Additionally, certain survey dates only included a total bird count for the day, but no location-level information. These records were excluded from the initial database query (24 survey dates out of a total of 538).

After excluding these surveys, the initial dataset queried from the `bppeeps` dataset included 1727 records from a total of 469. This is termed the ‘full dataset’ and was filtered further in R. The full dataset includes one survey per station per day.

Filtering step	No. survey dates	No. records	No. survey dates lost	No. records lost
Full dataset	469	1727	NA	NA
Remove NA count records	469	1722	0	-5
Exclude dates where total # of birds < 1000	443	1633	-26	-89
Exclude dates outside of survey period (<04-15 or >05-15)	430	1584	-13	-49
Exclude Intercauseway and NA stations (e.g. location was simply ‘inner mud’, ‘mumblies’, ‘flying’, etc.)	429	1477	-1	-107
Exclude records where only bird count occurs in location that spans >2 stations (e.g., ‘BP to CP’)	426	1471	-3	-6

2.1 dat data summary after filtering

```

##   survey_date      start_time          station_n
## Min.   :1997-04-21  Length:1471      Pilings       :331
## 1st Qu.:2004-04-26  Class  :character  Coal Port     :326
## Median :2009-04-28  Mode   :character  Brunswick Point:269
## Mean   :2009-09-28                    34th St pullout:189
## 3rd Qu.:2015-04-19                    View corner   :170
## Max.   :2022-05-04                    Bend          :111
##                                         (Other)        : 75
##
##   station_s    final_count      p_wesa      raptor_count
## Coal Port      : 48  Min.   : 0  Min.   : 0.00  Min.   :1.000
## View corner    : 40  1st Qu.: 1138  1st Qu.: 53.12  1st Qu.:1.000
## Bend          : 21  Median  : 6500  Median  : 80.66  Median  :1.000
## 34th St pullout: 17  Mean   : 16333  Mean   : 69.34  Mean   :1.812
## Pilings        : 12  3rd Qu.: 21250  3rd Qu.: 91.18  3rd Qu.:2.000
## (Other)        :  1  Max.   :222500  Max.   :100.00  Max.   :8.000
## NA's           :1332  NA's   :458   NA's   :818
##
##   elev_min      elev_max      elev_median      elev_mean
## Min.   :-0.040  Min.   :3.96  Min.   :2.240  Min.   :2.660
## 1st Qu.: 0.830  1st Qu.:4.30  1st Qu.:3.072  1st Qu.:2.971
## Median : 1.210  Median :4.43  Median :3.255  Median :3.038
## Mean   : 1.223  Mean   :4.42  Mean   :3.238  Mean   :3.043
## 3rd Qu.: 1.620  3rd Qu.:4.56  3rd Qu.:3.410  3rd Qu.:3.120
## Max.   : 2.420  Max.   :4.85  Max.   :3.735  Max.   :3.422
##
##   elev_range      flow      total_precip      mean_temp
## Min.   :1.870  Min.   : 996  Min.   : 0.00  Min.   : 3.80
## 1st Qu.:2.720  1st Qu.:2490  1st Qu.: 0.00  1st Qu.: 9.30
## Median :3.210  Median :3110  Median : 0.00  Median :10.50
## Mean   :3.197  Mean   :3291  Mean   : 2.11  Mean   :10.68
## 3rd Qu.:3.610  3rd Qu.:4070  3rd Qu.: 2.00  3rd Qu.:11.90
## Max.   :4.640  Max.   :7830  Max.   :28.20  Max.   :17.50
## NA's   :5      NA's   :4
##
##   u            v      windspd      wind_deg
## Min.   :-19.8067  Min.   :-42.3467  Min.   : 5.292  Min.   :-177.35
## 1st Qu.: -3.8203  1st Qu.: -9.7841  1st Qu.:10.521  1st Qu.: -77.27
## Median : -0.9551  Median : -0.4341  Median :12.958  Median : -24.09
## Mean   : -0.7154  Mean   : -1.7433  Mean   :14.309  Mean   : 14.30
## 3rd Qu.:  2.5179  3rd Qu.:  8.1678  3rd Qu.:16.896  3rd Qu.: 116.65
## Max.   : 15.0489  Max.   : 20.4541  Max.   :43.750  Max.   : 179.93
##
##   year      julian_day      dos.V1      n_s
## 2019   :106  Min.   :105.0  Min.   :-1.8553478  N:956
## 2012   :104  1st Qu.:112.0  1st Qu.:-0.8711370  S:515
## 2008   : 90  Median :118.0  Median :-0.0275277
## 2001   : 80  Mean   :118.2  Mean   : 0.0000000
## 2005   : 80  3rd Qu.:124.0  3rd Qu.: 0.8160816
## 2015   : 70  Max.   :135.0  Max.   : 2.3626986
## (Other):941

```

3 Extended Canham et al. (2021) model

In this initial analysis, the model is kept relatively simple: the exact same model as in Canham et al. (2021) is used, but a “north vs. south” term is added. The response variable will be the log-transformed count of WESA for that location (total count for that location/day * the predicted WESA ratio generated by the bGLMM above).

3.0.1 Base model

Response variable:

- `log_wesa` - Log-transformed predicted WESA count

Predictor variables:

- `year_c` - Scale-transformed survey year
- `dos` - Scale-transformed Julian date, aka Day of Season

```
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [  
## lmerModLmerTest]  
## Formula: log_wesa ~ year_c + dos + I(dos^2) + (dos + I(dos^2) | year)  
##   Data: dat  
##  
## REML criterion at convergence: 7302.2  
##  
## Scaled residuals:  
##      Min       1Q     Median       3Q      Max  
## -3.2987 -0.3248  0.2468  0.6247  2.1576  
##  
## Random effects:  
##   Groups   Name        Variance Std.Dev. Corr  
##   year     (Intercept) 0.3804   0.6168  
##           dos         0.5523   0.7432  -0.12  
##           I(dos^2)   0.1273   0.3568  -0.13 -0.04  
##   Residual    7.9024   2.8111  
## Number of obs: 1471, groups: year, 24  
##  
## Fixed effects:  
##             Estimate Std. Error      df t value Pr(>|t|)  
## (Intercept)  8.57114   0.16581 15.94079  51.693 < 2e-16 ***  
## year_c     -0.02345   0.14500 22.31502  -0.162   0.873  
## dos        -0.04050   0.17635 19.23766  -0.230   0.821  
## I(dos^2)   -1.32114   0.11598 11.08207 -11.391 1.85e-07 ***  
## ---  
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## Correlation of Fixed Effects:  
##          (Intr) year_c dos  
## year_c   -0.014  
## dos      -0.078  0.017  
## I(dos^2) -0.398  0.026 -0.007
```

3.0.2 Full model

NOTE: at the time of this writing, IR is *unavailable* in this model. I have not gotten any response from the team that manages the UBC Totem Station data. Given it was not a significant variable in Canham et al. (2021), I do not believe it is urgent to acquire this particular data.

Response variable:

- `log_wesa` - Log-transformed predicted WESA count

Predictor variables:

- `year_c` - Scale-transformed survey year
- `dos` - Scale-transformed Julian date, aka Day of Season
- `elev_range` - Tidal amplitude (m)
- `total_precip` - Total daily precipitation (mm)
- `mean_temp` - Daily mean temperature (C°)
- `flow` - Fraser River discharge (m^3/s)
- `u, v` - Westerly and Southerly wind vectors (km/h)
- `n_s` - Location ('North' or 'South')

Additionally, the dataset is reduced down to only complete cases of all predictor variables of interest (1471 -> 1462).

```
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [  
## lmerModLmerTest]  
## Formula:  
## log_wesa ~ year_c + dos + I(dos^2) + scale(elev_range) + scale(total_precip) +  
##     scale(mean_temp) + scale(flow) + scale(u) + scale(v) + n_s +  
##     (dos + I(dos^2) | year) + (scale(flow) | n_s)  
## Data: dat2  
##  
## REML criterion at convergence: 7182.4  
##  
## Scaled residuals:  
##      Min       1Q   Median       3Q      Max  
## -3.4130 -0.3223  0.2407  0.6394  2.1634  
##  
## Random effects:  
## Groups   Name        Variance Std.Dev. Corr  
## year     (Intercept) 4.460e-01 6.679e-01  
##           dos         6.103e-01 7.812e-01 -0.22  
##           I(dos^2)   6.387e-02 2.527e-01  0.46 -0.04  
## n_s      (Intercept) 6.145e+00 2.479e+00  
##           scale(flow) 1.082e-09 3.289e-05 1.00  
## Residual    7.421e+00 2.724e+00  
## Number of obs: 1462, groups: year, 24; n_s, 2  
##  
## Fixed effects:  
##             Estimate Std. Error      df t value Pr(>|t|)  
## (Intercept) 9.10629  2.48556 1383.39733  3.664 0.000258 ***  
## year_c     -0.11093  0.16058  20.78506 -0.691 0.497335  
## dos        0.08759  0.20620  26.92175  0.425 0.674382
```

```

## I(dos^2)           -1.38697   0.10097   11.70827 -13.737 1.41e-08 ***
## scale(elev_range)  0.25045   0.08231   484.67379   3.043 0.002471 **
## scale(total_precip) 0.03184   0.08023  1297.14887   0.397 0.691563
## scale(mean_temp)    0.13624   0.09490   694.28100   1.436 0.151569
## scale(flow)         -0.24502   0.14872   41.26802  -1.648 0.107043
## scale(u)            0.08265   0.09278   825.77108   0.891 0.373329
## scale(v)            0.23498   0.09595  1091.10794   2.449 0.014485 *
## n_ss                -1.40193   3.50904  1369.17555  -0.400 0.689573
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##          (Intr) year_c dos   I(d^2) scl(l_) scl(t_) scl(m_) scl(f) scl()
## year_c     -0.001
## dos        -0.006 -0.012
## I(dos^2)   -0.012  0.032 -0.074
## scl(lv_rng) 0.000  0.025 -0.039  0.015
## scl(ttl_pr) 0.000  0.024 -0.011  0.026  0.045
## scl(mn_tmp) -0.001 -0.042 -0.149  0.007 -0.081   0.064
## scale(flow)  -0.007  0.041 -0.444  0.135  0.050   0.043 -0.009
## scale(u)      0.003 -0.056  0.044 -0.075 -0.089  -0.038   0.062 -0.004
## scale(v)      0.002 -0.011  0.063 -0.069  0.039  -0.260  -0.079  -0.035  0.514
## n_ss        -0.706  0.000  0.000  0.000 -0.002   0.000   0.001  -0.001  0.000
##               scl(v)
## year_c
## dos
## I(dos^2)
## scl(lv_rng)
## scl(ttl_pr)
## scl(mn_tmp)
## scale(flow)
## scale(u)
## scale(v)
## n_ss       0.000
## optimizer (nloptwrap) convergence code: 0 (OK)
## boundary (singular) fit: see help('isSingular')

```

3.0.3 Backwards step-wise selection

```

## Backward reduced random-effect table:
##
##                                         Eliminated npar logLik AIC LRT Df
## <none>                                21 -3591.2 7224.4
## dos in (dos + I(dos^2) | year)          0   18 -3611.2 7258.4 40.033  3
## I(dos^2) in (dos + I(dos^2) | year)      0   18 -3592.9 7221.8  3.355  3
## scale(flow) in (scale(flow) | n_s)        0   19 -3591.2 7220.4  0.000  2
##                                         Pr(>Chisq)
## <none>
## dos in (dos + I(dos^2) | year)      1.048e-08 ***
## I(dos^2) in (dos + I(dos^2) | year)    0.34
## scale(flow) in (scale(flow) | n_s)      1.00
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```

## 
## Backward reduced fixed-effect table:
## Degrees of freedom method: Satterthwaite
## 
##          Eliminated   Sum Sq Mean Sq NumDF DenDF F value    Pr(>F)
## scale(total_precip)      1   1.17   1.17     1 1297.15  0.1575 0.691563
## n_s                      2   0.69   0.69     1 1370.49  0.0928 0.760725
## scale(u)                  3   6.14   6.14     1  823.09  0.8279 0.363157
## scale(mean_temp)         4  13.67  13.67     1  711.85  1.8445 0.174857
## scale(flow)                5  21.38  21.38     1   18.61  2.8839 0.106132
## year_c                     0   0.50   0.50     1   21.79  0.0671 0.798097
## dos                       0   0.30   0.30     1   19.63  0.0403 0.842911
## I(dos^2)                   0 1066.28 1066.28     1   12.04 144.0700 4.663e-08
## scale(elev_range)          0   80.14  80.14     1   688.01 10.8282 0.001051
## scale(v)                    0   49.87  49.87     1 1189.52  6.7387 0.009551
## 
## scale(total_precip)
## n_s
## scale(u)
## scale(mean_temp)
## scale(flow)
## year_c
## dos
## I(dos^2)           ***
## scale(elev_range)  **
## scale(v)            **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## 
## Model found:
## log_wesa ~ year_c + dos + I(dos^2) + scale(elev_range) + scale(v) + (dos + I(dos^2) | year) + (scale

```

3.0.4 Summary of best-fit model

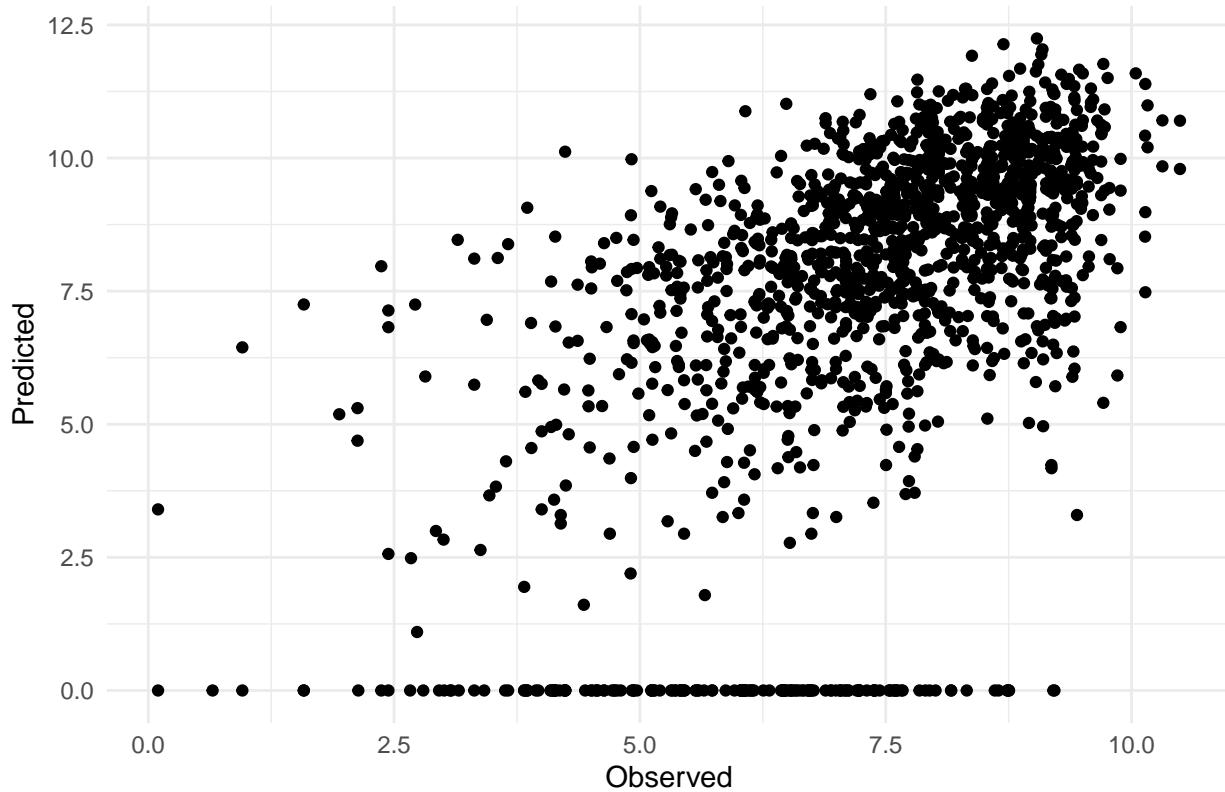
```

## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
## Formula: log_wesa ~ year_c + dos + I(dos^2) + scale(elev_range) + scale(v) +
##           (dos + I(dos^2) | year) + (scale(flow) | n_s)
## Data: dat2
##
## REML criterion at convergence: 7179.9
##
## Scaled residuals:
##      Min     1Q Median     3Q    Max
## -3.3876 -0.3149  0.2521  0.6390  2.1610
##
## Random effects:
##   Groups   Name        Variance Std.Dev. Corr
##   year     (Intercept) 0.472474 0.68737
##           dos          0.634054 0.79628 -0.16
##           I(dos^2)    0.116392 0.34116  0.03  0.02
##   n_s      (Intercept) 1.013085 1.00652
##           scale(flow) 0.001794 0.04235  1.00
##   Residual       7.401143 2.72050
## Number of obs: 1462, groups: year, 24; n_s, 2
##
## Fixed effects:
##             Estimate Std. Error      df t value Pr(>|t|)    
## (Intercept)  8.55731  0.72403  1.11667 11.819 0.04136 *
## year_c      -0.04111  0.15874  21.79071 -0.259 0.79810
## dos         -0.03736  0.18606  19.63071 -0.201 0.84291
## I(dos^2)    -1.34797  0.11230  12.03937 -12.003 4.66e-08 ***
## scale(elev_range) 0.27225  0.08273 688.00994  3.291 0.00105 **
## scale(v)      0.20483  0.07890 1189.52245  2.596 0.00955 ** 
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##            (Intr) year_c dos   I(d^2) scl(_)
## year_c     -0.008
## dos        0.069  0.008
## I(dos^2)   -0.095  0.024  0.020
## scl(lv_rng) -0.013  0.011 -0.033  0.013
## scale(v)    0.004  0.019  0.015 -0.027  0.110
## optimizer (nloptwrap) convergence code: 0 (OK)
## boundary (singular) fit: see help('isSingular')

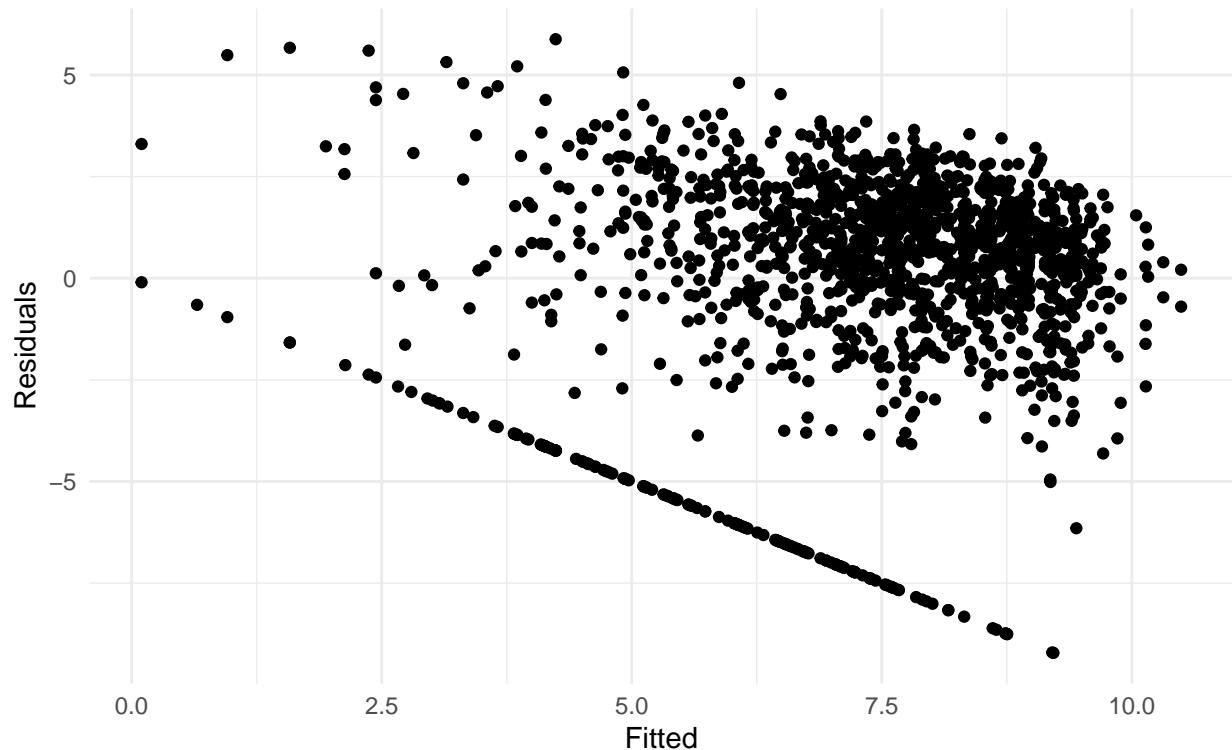
```

3.0.5 Check assumptions of best fit model

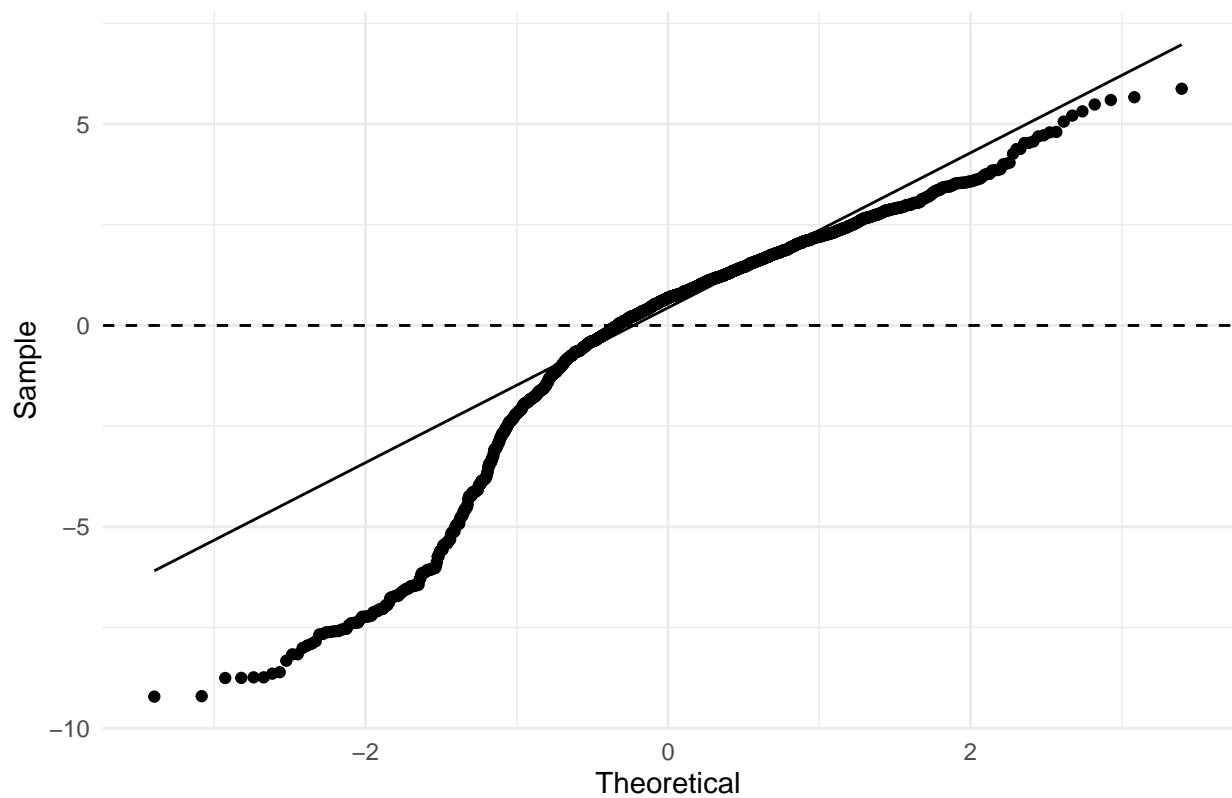
Observed vs. Predicted values



Heteroskedasticity
Fitted values vs. Residuals



Quantile–Quantile



4 Exploring mechanism (flow) vs. time (year)

Given the Canham extended model has some issues, the next step is to run two models with a slightly simpler approach. We are still interested in seeing if the discharge (`flow`) has an impact on the north-south distribution of peeps. In addition, we are interested in seeing any temporal changes in this distribution.

Two simpler models can get at these questions:

1. The ‘mechanism’ model: `log_wesa ~ n_s * scale(flow) + dos + I(dos^2) + (dos + I(dos^2) | year)`
2. The ‘temporal’ model: `log_wesa ~ n_s * year_c + dos + I(dos^2) + (dos + I(dos^2) | year)`

4.1 Data changes

These simpler models necessitate further data aggregation. We are primarily interested in N vs S distribution of peeps: therefore, the data are grouped by N/S such that there are two data points per survey date: number of peeps in the northern stations (Canoe Pass, Brunswick Point, Bend, Pilings, View corner) vs southern (34th St pullout, Coal Port). Any NA values in our variables of interest are also removed.

```
##   year survey_date julian_day      dos start_time n_s final_count
## 1 1997 1997-04-21       111 -1.0204526 17:35     N    14030
## 2 1997 1997-04-21       111 -1.0204526 18:05     S      0
## 3 1997 1997-04-22       112 -0.8808935 00:00     N   13000
## 4 1997 1997-04-22       112 -0.8808935 17:25     S      0
## 5 1997 1997-04-25       115 -0.4622159 09:15     N   76100
## 6 1997 1997-04-25       115 -0.4622159 09:00     S   54000
##   predicted_wesa predicted_dunl log_wesa log_dunl p_wesa predicted_ratio
## 1          3506        10524  8.162516 9.261509 60.000000  0.2499045
## 2            0           0  0.000000 0.000000 60.000000  0.2499045
## 3          5328        7672  8.580919 8.945463 52.31788  0.4098783
## 4            0           0  0.000000 0.000000 52.31788  0.4098783
## 5          61335       14765 11.024122 9.600083 70.58319  0.8059765
## 6          43523       10477 10.681068 9.257033 70.58319  0.8059765
##   raptor_count elev_min elev_max elev_median elev_mean elev_range flow
## 1             1    1.60    4.13    3.170  3.056667    2.53 3430
## 2             1    1.60    4.13    3.170  3.056667    2.53 3430
## 3            NA    1.42    4.28    3.235  3.109167    2.86 3670
## 4            NA    1.42    4.28    3.235  3.109167    2.86 3670
## 5            NA    0.89    4.34    3.155  3.003750    3.45 4160
## 6            NA    0.89    4.34    3.155  3.003750    3.45 4160
##   total_precip mean_temp          u          v windsspd wind_deg year_n
## 1         0.0     8.3  2.952019 -6.894400 12.45833 -37.88846     1
## 2         0.0     8.3  2.952019 -6.894400 12.45833 -37.88846     1
## 3         5.6    10.7 -2.325784  3.963163 11.16667 147.32756     1
## 4         5.6    10.7 -2.325784  3.963163 11.16667 147.32756     1
## 5         3.4    12.7 -1.041889  7.000000  9.25000  92.45501     1
## 6         3.4    12.7 -1.041889  7.000000  9.25000  92.45501     1
##   year_c
## 1 -1.663883
## 2 -1.663883
## 3 -1.663883
## 4 -1.663883
```

```

## 5 -1.663883
## 6 -1.663883

##      year      survey_date      julian_day      dos.V1
## 2019 : 51    Min. :1997-04-21    Min. :105.0   Min. :-1.8578077
## 2012 : 46    1st Qu.:2003-05-04  1st Qu.:112.0   1st Qu.:-0.8808935
## 2008 : 45    Median :2008-05-09  Median :118.0   Median :-0.0435384
## 2015 : 45    Mean   :2009-03-11  Mean   :118.3   Mean   :-0.0008058
## 2005 : 44    3rd Qu.:2014-10-24  3rd Qu.:124.0   3rd Qu.: 0.7938166
## 1999 : 42    Max.   :2022-05-04  Max.   :135.0   Max.   : 2.3289675
## (Other):550

##      start_time      n_s      final_count      predicted_wesa      predicted_dunl
## Length:823      N:419      Min. : 0      Min. : 0      Min. : 0
## Class :character  S:404      1st Qu.: 4000  1st Qu.: 2298  1st Qu.: 557
## Mode  :character                    Median :16100   Median : 9116   Median : 2752
##                                         Mean   :29125   Mean   :20961   Mean   : 8163
##                                         3rd Qu.:43000  3rd Qu.:27900  3rd Qu.: 9560
##                                         Max.   :223500  Max.   :208942  Max.   :100284
##
##      log_wesa      log_dunl      p_wesa      predicted_ratio
## Min. : 0.000  Min. : 0.000  Min. : 0.00  Min. :0.008502
## 1st Qu.: 7.740 1st Qu.: 6.324 1st Qu.: 53.04 1st Qu.:0.577486
## Median : 9.118 Median : 7.920 Median : 81.08 Median :0.833322
## Mean   : 8.284 Mean   : 7.223 Mean   : 69.25 Mean   :0.707910
## 3rd Qu.:10.236 3rd Qu.: 9.165 3rd Qu.: 91.49 3rd Qu.:0.908682
## Max.   :12.250 Max.   :11.516 Max.   :100.00 Max.   :0.985180
## NA's   :277

##      raptor_count      elev_min      elev_max      elev_median      elev_mean
## Min. :1.0  Min. :-0.040  Min. :3.96  Min. :2.240  Min. :2.660
## 1st Qu.:1.0 1st Qu.: 0.865 1st Qu.:4.30  1st Qu.:3.080  1st Qu.:2.973
## Median :1.0  Median : 1.240  Median :4.43  Median :3.255  Median :3.041
## Mean   :1.8  Mean   : 1.235  Mean   :4.42  Mean   :3.240  Mean   :3.047
## 3rd Qu.:2.0 3rd Qu.: 1.640  3rd Qu.:4.56  3rd Qu.:3.405  3rd Qu.:3.121
## Max.   :8.0  Max.   : 2.420  Max.   :4.85  Max.   :3.735  Max.   :3.422
## NA's   :457

##      elev_range      flow      total_precip      mean_temp
## Min. :1.870  Min. : 996  Min. : 0.000  Min. : 3.80
## 1st Qu.:2.710 1st Qu.:2520 1st Qu.: 0.000  1st Qu.: 9.30
## Median :3.190 Median :3140  Median : 0.000  Median :10.50
## Mean   :3.185 Mean   :3322  Mean   : 2.069  Mean   :10.71
## 3rd Qu.:3.600 3rd Qu.:4060 3rd Qu.: 2.000  3rd Qu.:12.00
## Max.   :4.640 Max.   :7830  Max.   :28.200  Max.   :17.50
##
##      u          v      windspd      wind_deg
## Min. :-19.807  Min. :-42.347  Min. : 5.292  Min. :-177.35
## 1st Qu.:-4.104 1st Qu.:-9.642  1st Qu.:10.625 1st Qu.:-76.58
## Median :-1.042 Median : 0.000  Median :13.000  Median :-10.15
## Mean   :-0.802 Mean   :-1.522  Mean   :14.249  Mean   : 15.34
## 3rd Qu. :2.367 3rd Qu.: 8.391  3rd Qu.:16.729 3rd Qu.:118.55
## Max.   :15.049 Max.   :20.454  Max.   :43.750  Max.   :179.93
##
##      year_n      year_c.V1
## Min. : 1.00  Min. :-1.6638826
## 1st Qu.: 6.00 1st Qu.:-0.9010314

```

```

## Median :11.00  Median :-0.1381803
## Mean   :11.86  Mean   :-0.0063730
## 3rd Qu.:17.50  3rd Qu.: 0.8535262
## Max.   :24.00  Max.   : 1.8452327
##

```

4.2 The ‘mechanism’ model

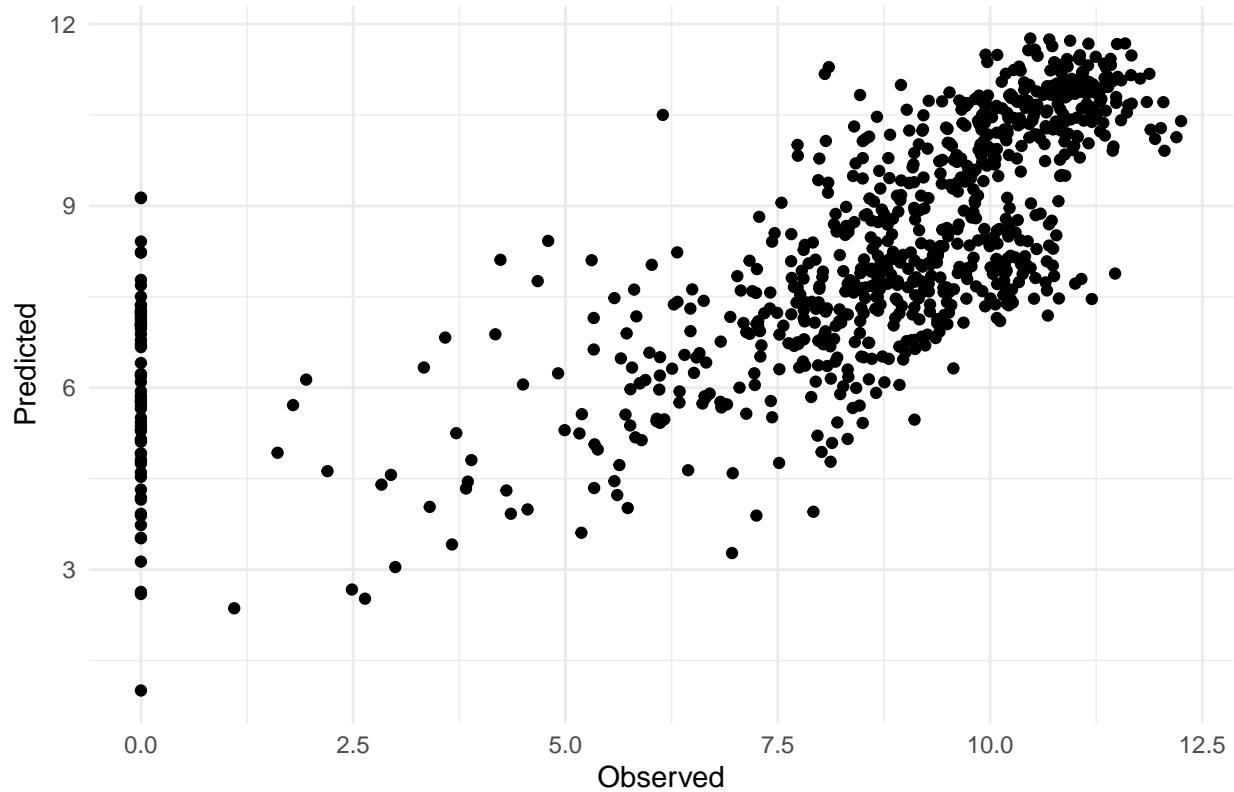
```

## Linear mixed model fit by REML ['lmerMod']
## Formula: log_wesa ~ n_s * scale(flow) + dos + I(dos^2) + (dos + I(dos^2) |
##           year)
## Data: dat3
##
## REML criterion at convergence: 3744.4
##
## Scaled residuals:
##    Min     1Q  Median     3Q    Max
## -4.0428 -0.2276  0.1400  0.5718  1.7551
##
## Random effects:
## Groups   Name        Variance Std.Dev. Corr
## year     (Intercept) 0.26220  0.5121
##          dos         0.47129  0.6865  -0.31
##          I(dos^2)    0.04913  0.2217  -0.08  0.59
## Residual            5.10678  2.2598
## Number of obs: 823, groups: year, 24
##
## Fixed effects:
##                   Estimate Std. Error t value
## (Intercept)      10.95996  0.17388 63.030
## n_sS             -2.91429  0.15768 -18.482
## scale(flow)      -0.24031  0.16048 -1.497
## dos              0.10666  0.18943  0.563
## I(dos^2)        -1.37393  0.10280 -13.365
## n_sS:scale(flow) -0.06884  0.15777 -0.436
##
## Correlation of Fixed Effects:
##          (Intr) n_sS scl(f) dos   I(d^2)
## n_sS      -0.444
## scale(flow) -0.086  0.003
## dos       -0.098 -0.002 -0.406
## I(dos^2)  -0.426 -0.002  0.113  0.167
## n_sS:scl(f) -0.002  0.000 -0.481 -0.001  0.009

```

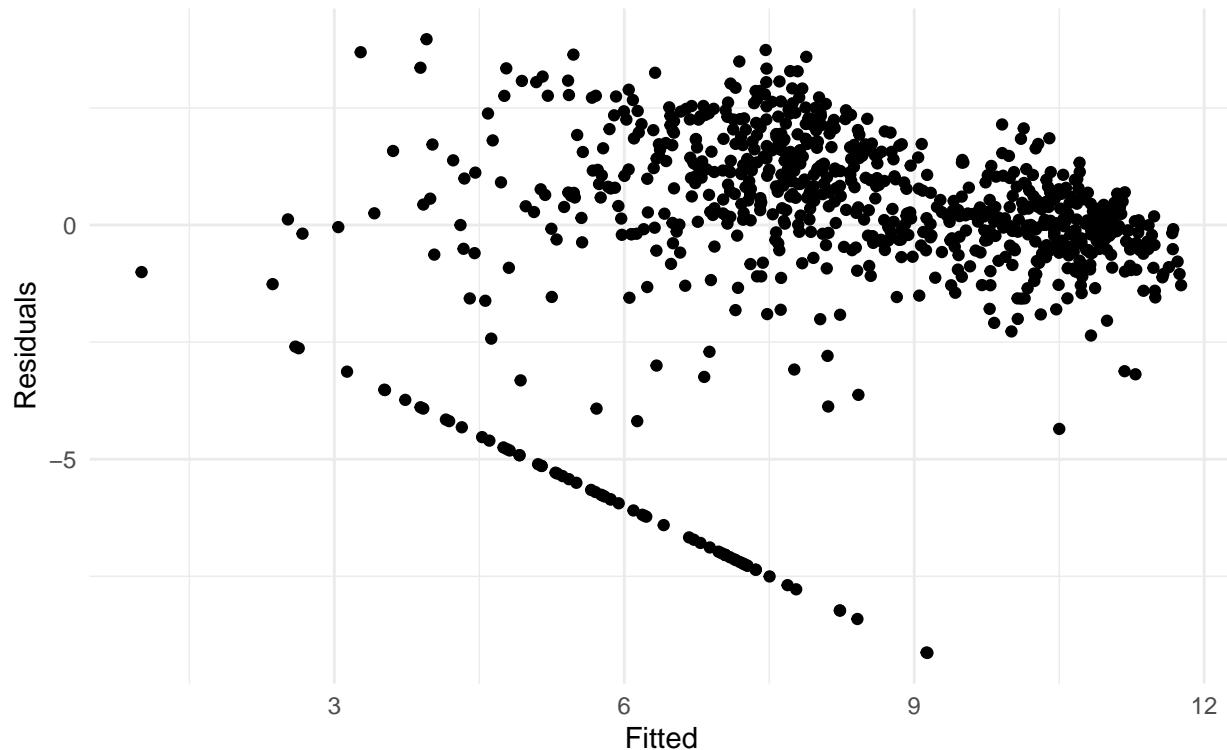
4.2.1 Diagnostic plots

Observed vs. Fitted values

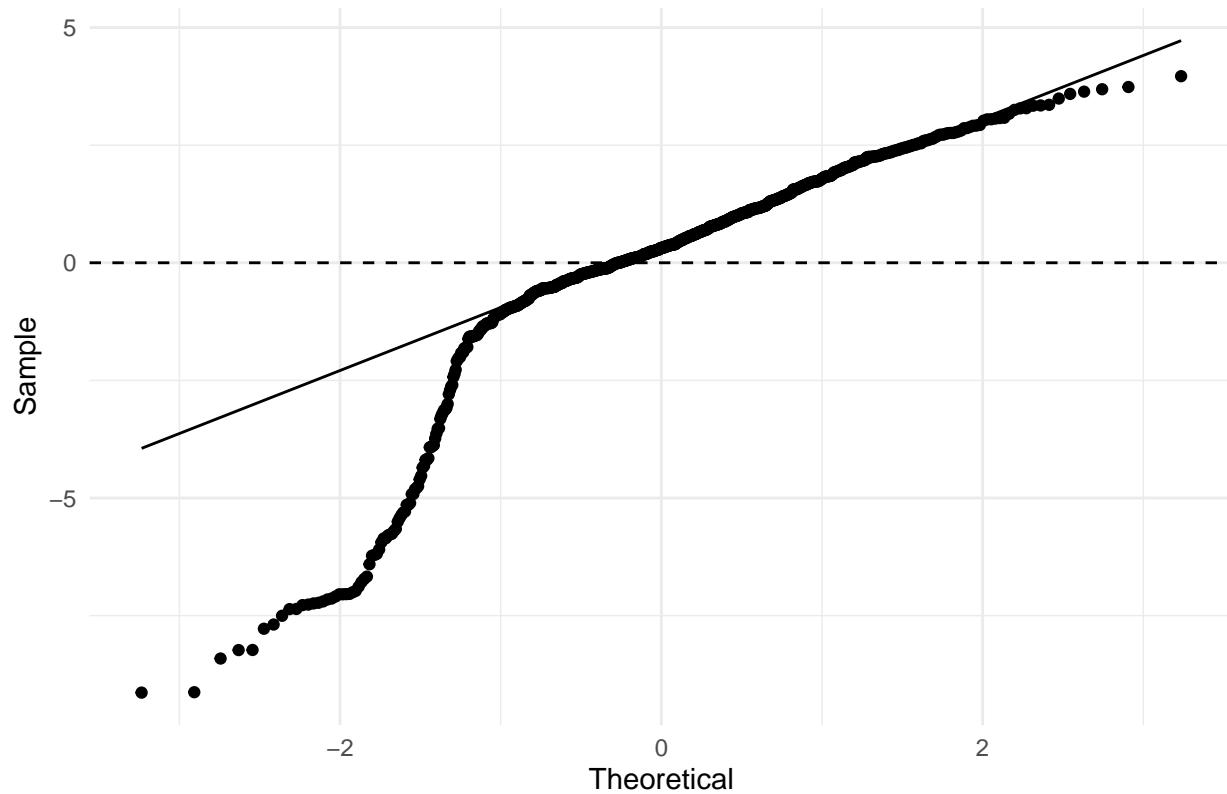


Heteroskedasticity

Fitted values vs. Residuals



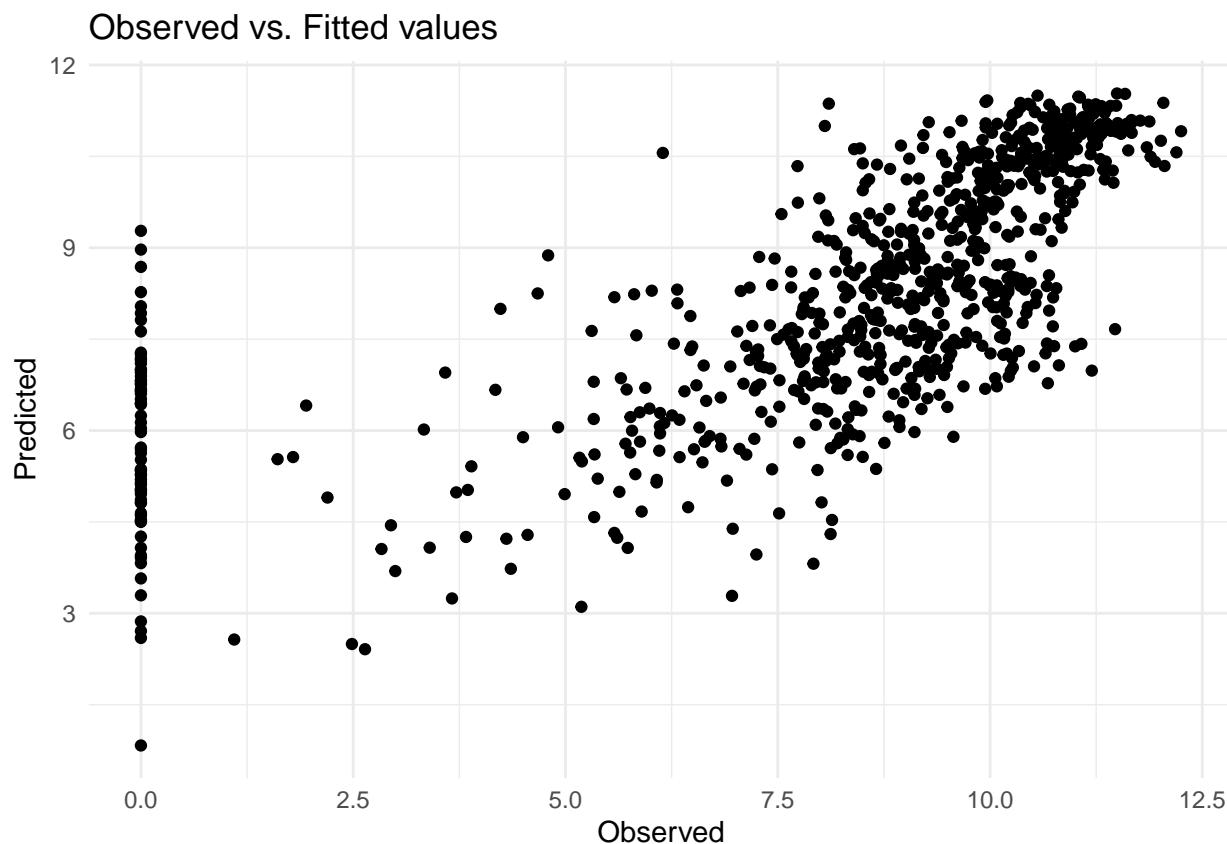
Quantile–Quantile



4.3 The ‘temporal’ model

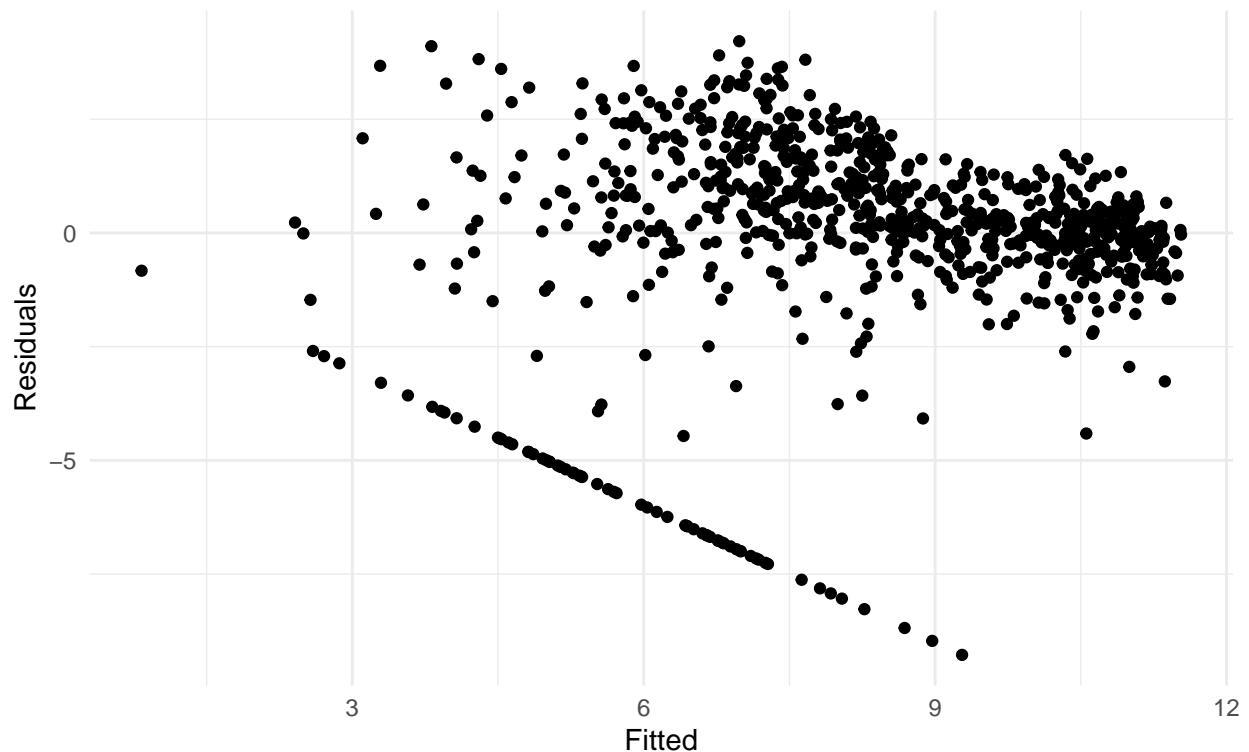
```
## Linear mixed model fit by REML ['lmerMod']
## Formula: log_wesa ~ n_s * year_c + dos + I(dos^2) + (dos + I(dos^2) | 
##   year)
## Data: dat3
##
## REML criterion at convergence: 3733.6
##
## Scaled residuals:
##    Min     1Q Median     3Q    Max
## -4.1323 -0.2056  0.1068  0.5320  1.8776
##
## Random effects:
##   Groups   Name        Variance Std.Dev. Corr
##   year     (Intercept) 0.25123  0.5012
##           dos          0.46676  0.6832  -0.02
##           I(dos^2)    0.07833  0.2799  -0.29  0.56
##   Residual            5.04042  2.2451
## Number of obs: 823, groups: year, 24
##
## Fixed effects:
##             Estimate Std. Error t value
## (Intercept) 10.91398  0.17126 63.727
## n_sS        -2.90810  0.15666 -18.563
## year_c      -0.04480  0.14707 -0.305
## dos         -0.07831  0.16730 -0.468
## I(dos^2)    -1.33924  0.10807 -12.393
## n_sS:year_c  0.56111  0.15710  3.572
##
## Correlation of Fixed Effects:
## (Intr) n_sS  year_c dos  I(d^2)
## n_sS  -0.448
## year_c -0.011 -0.003
## dos    -0.021  0.000  0.031
## I(dos^2) -0.472 -0.003 -0.010  0.299
## n_sS:year_c -0.005  0.007 -0.524 -0.004  0.004
```

4.3.1 Diagnostic plots

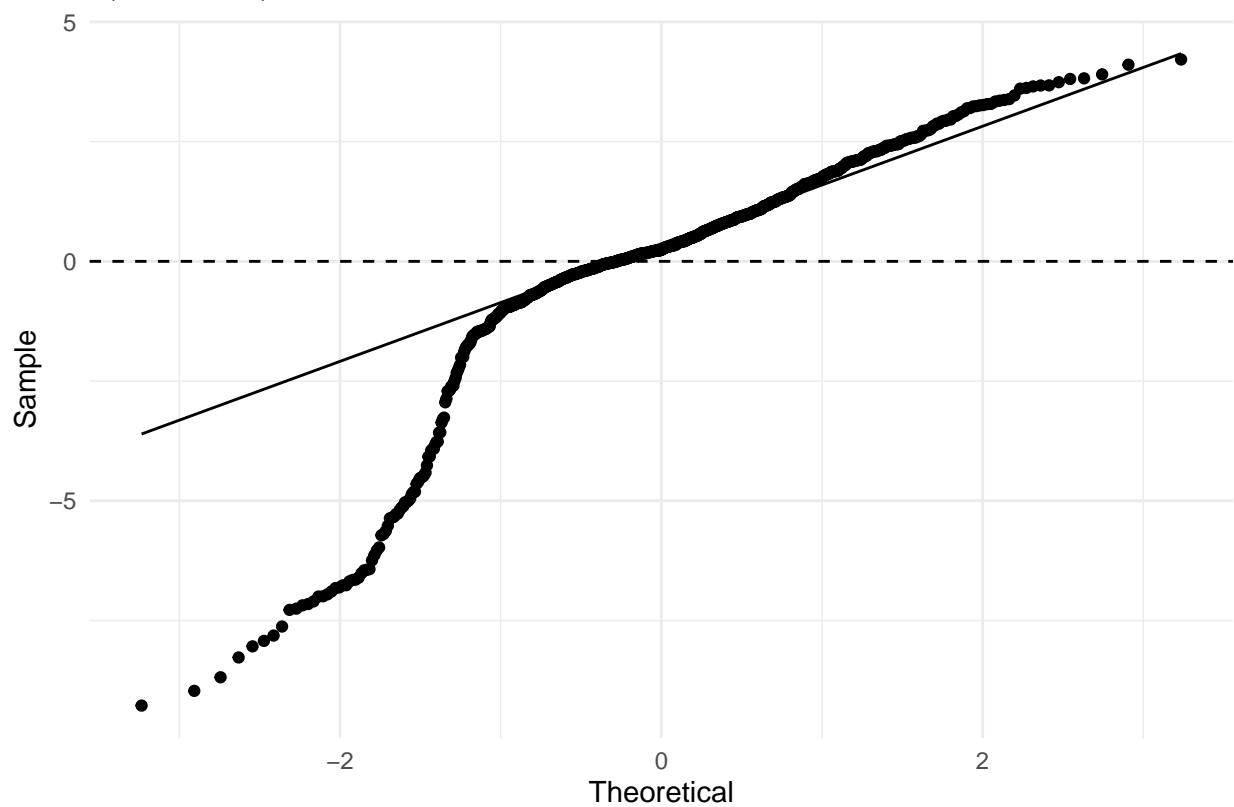


Heteroskedasticity

Fitted values vs. Residuals



Quantile–Quantile

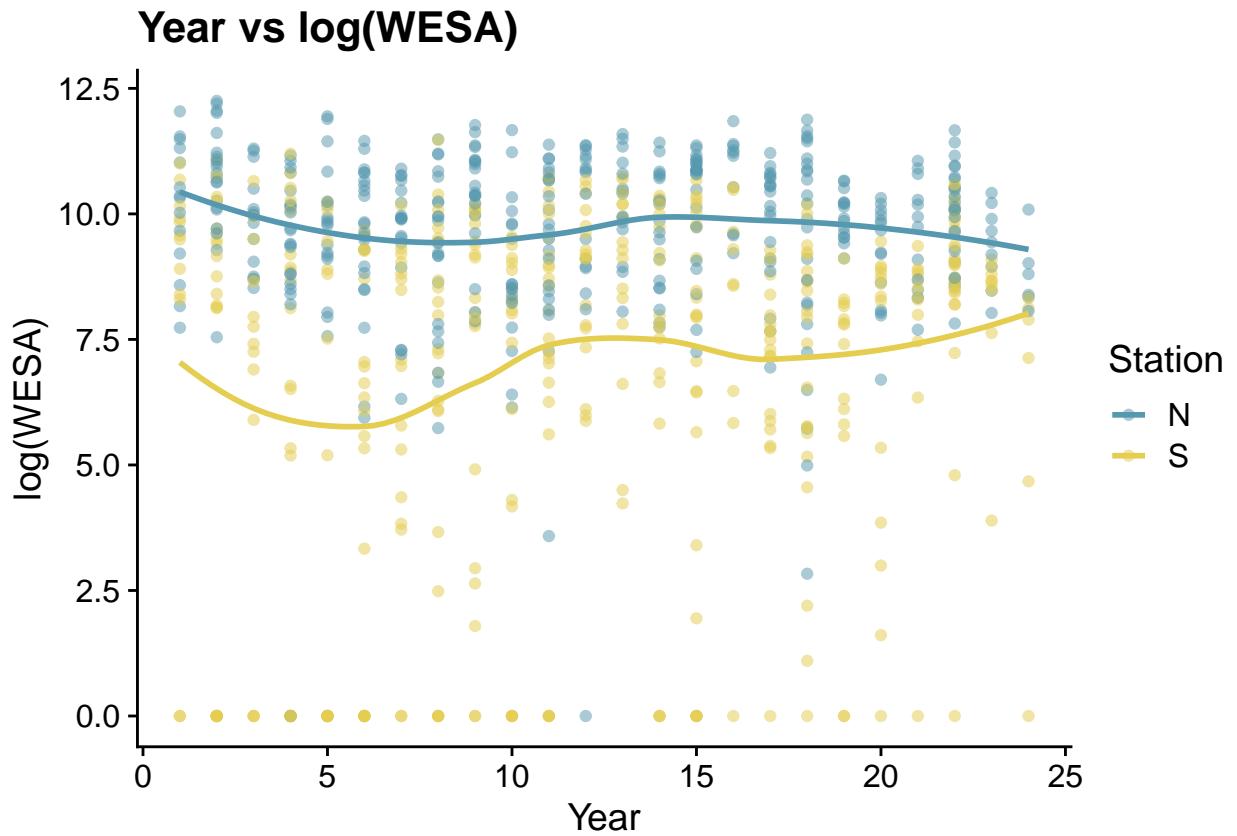


4.4 Comparing the two

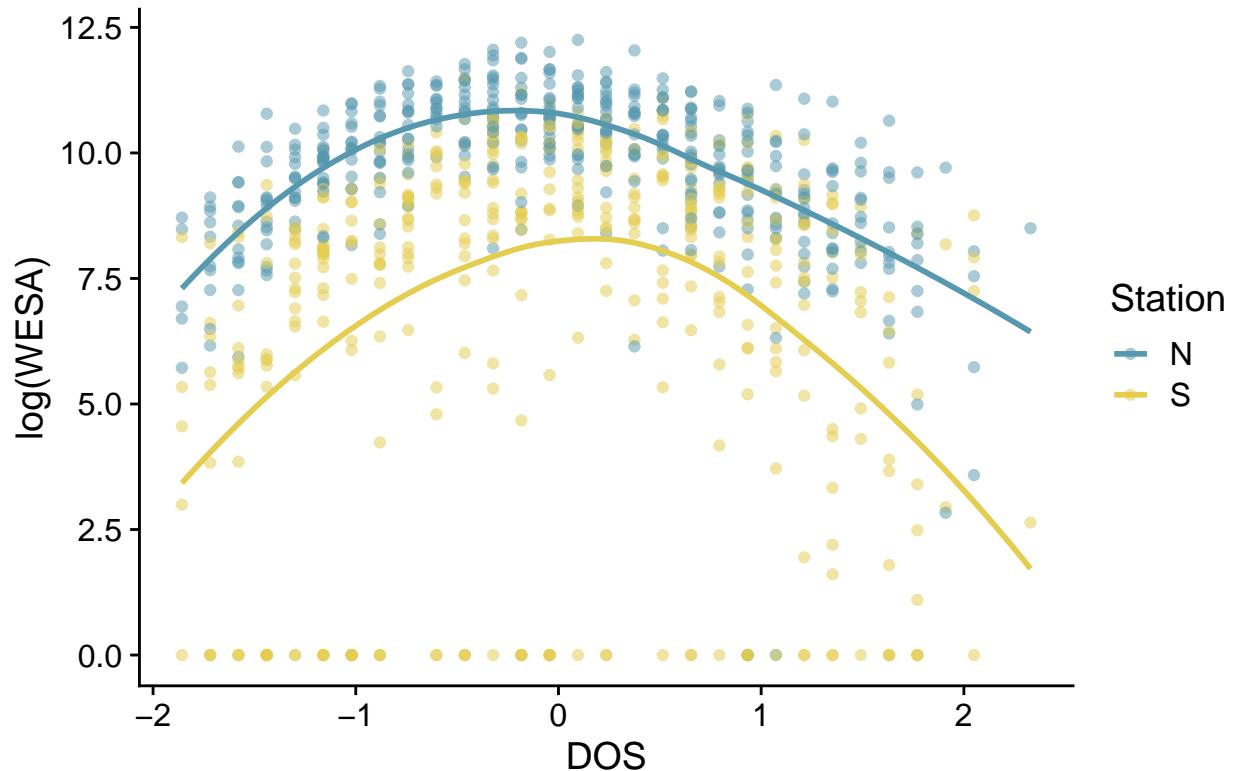
```
## Data: dat3
## Models:
## mech_model: log_wesa ~ n_s * scale(flow) + dos + I(dos^2) + (dos + I(dos^2) | year)
## temp_model: log_wesa ~ n_s * year_c + dos + I(dos^2) + (dos + I(dos^2) | year)
##          npar   AIC   BIC  logLik deviance Chisq Df Pr(>Chisq)
## mech_model    13 3757.7 3819.0 -1865.9   3731.7
## temp_model    13 3746.6 3807.9 -1860.3   3720.6 11.108  0
```

5 Final WESA model

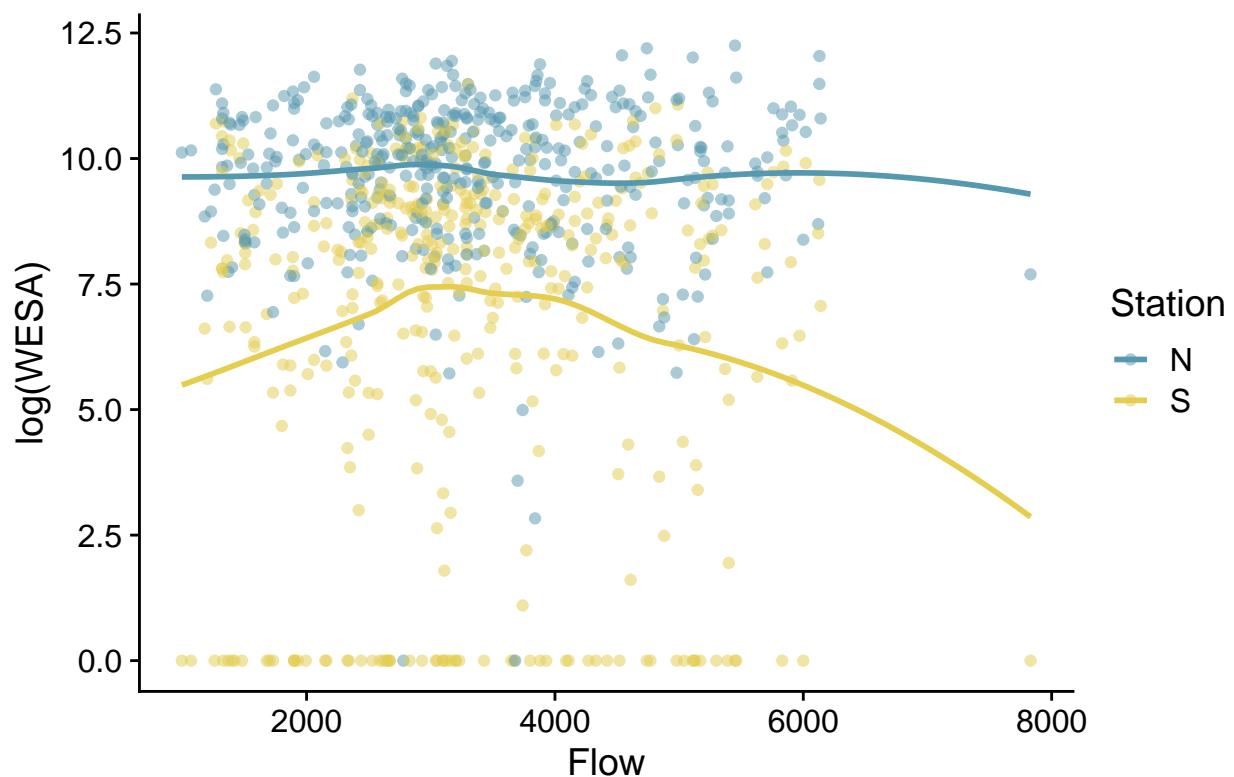
5.1 WESA vs. variable plots



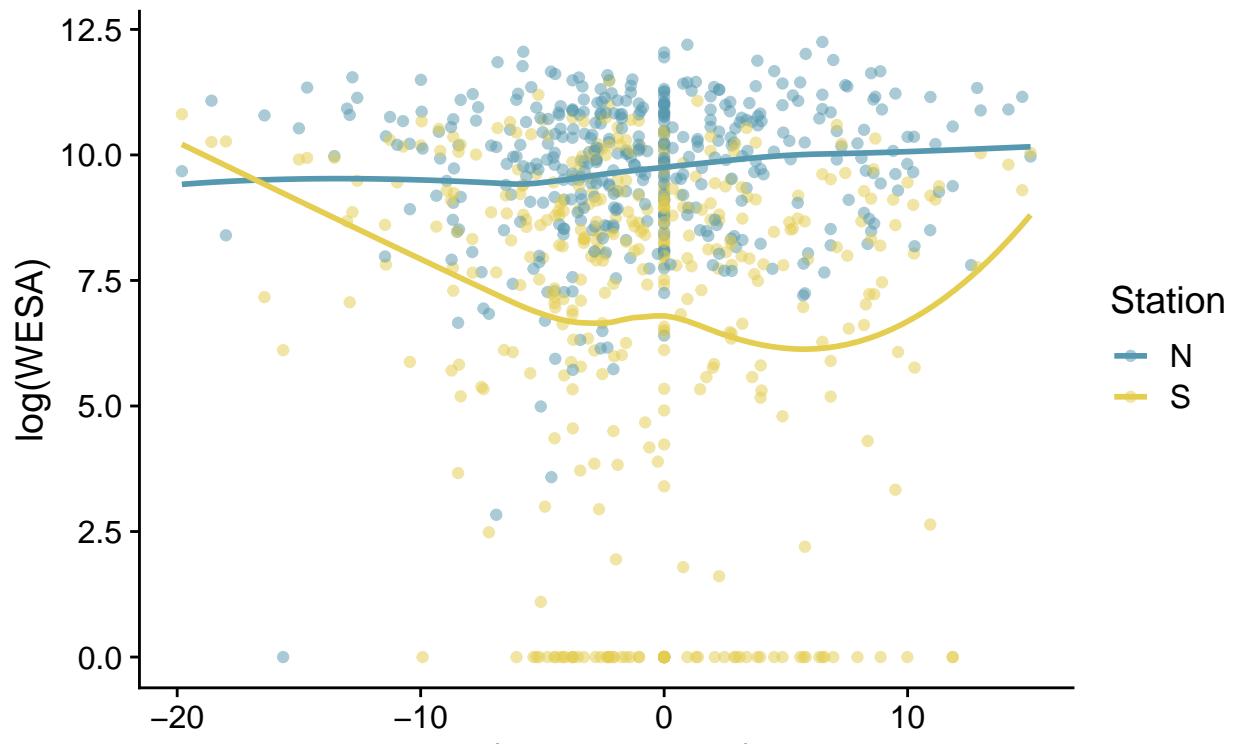
Day of Season vs log(WESA)



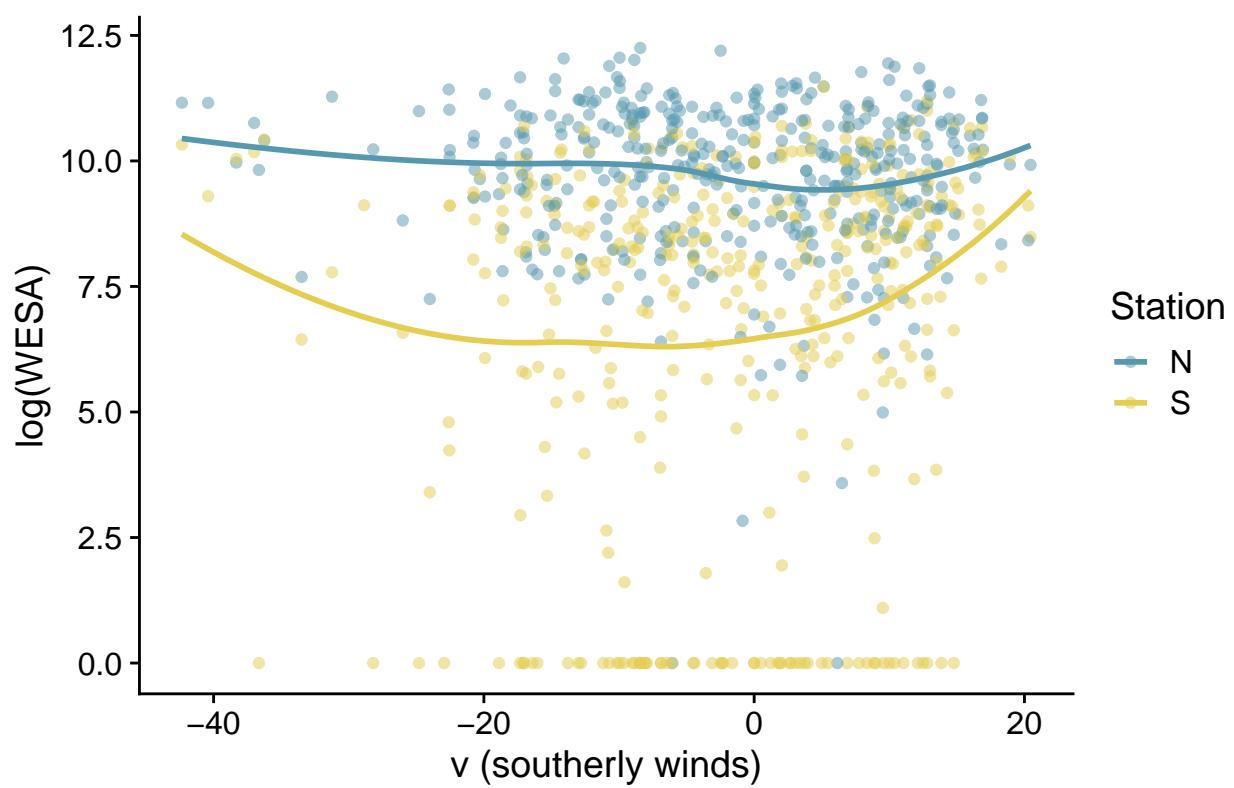
Flow vs log(WESA)



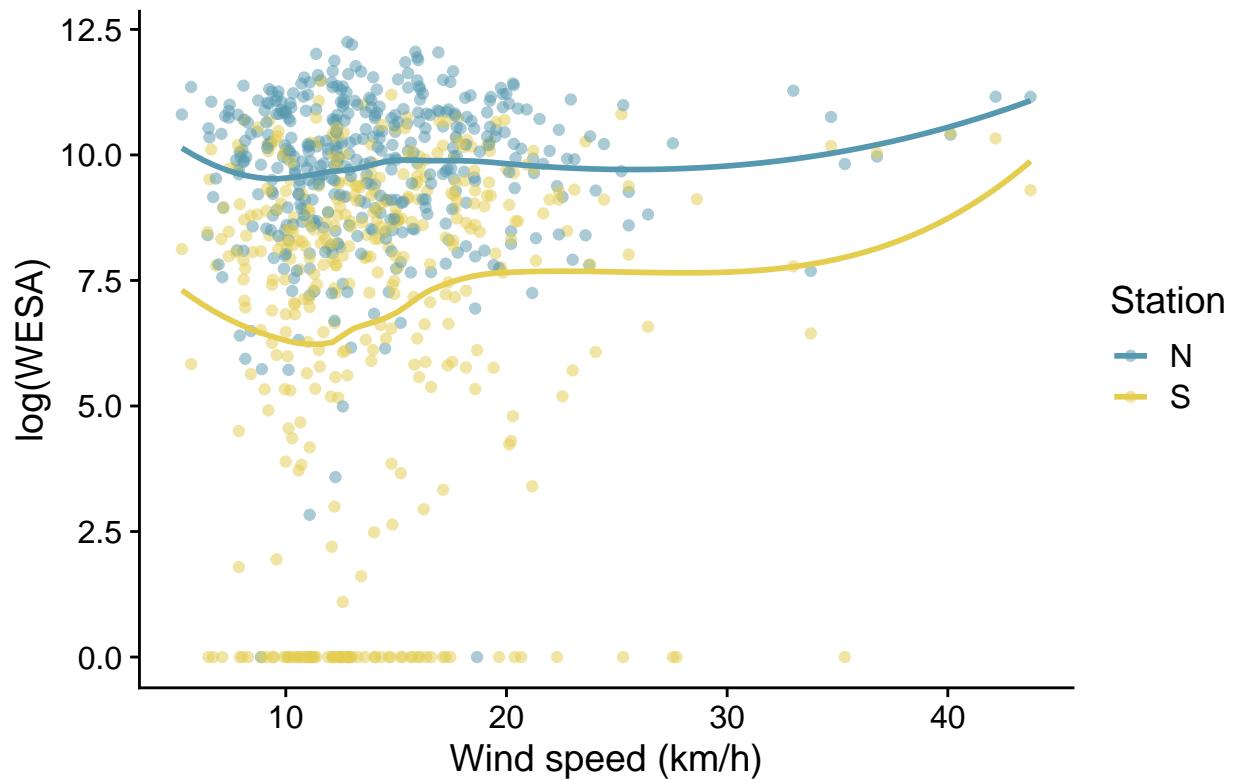
u (westerly winds) vs log(WESA)



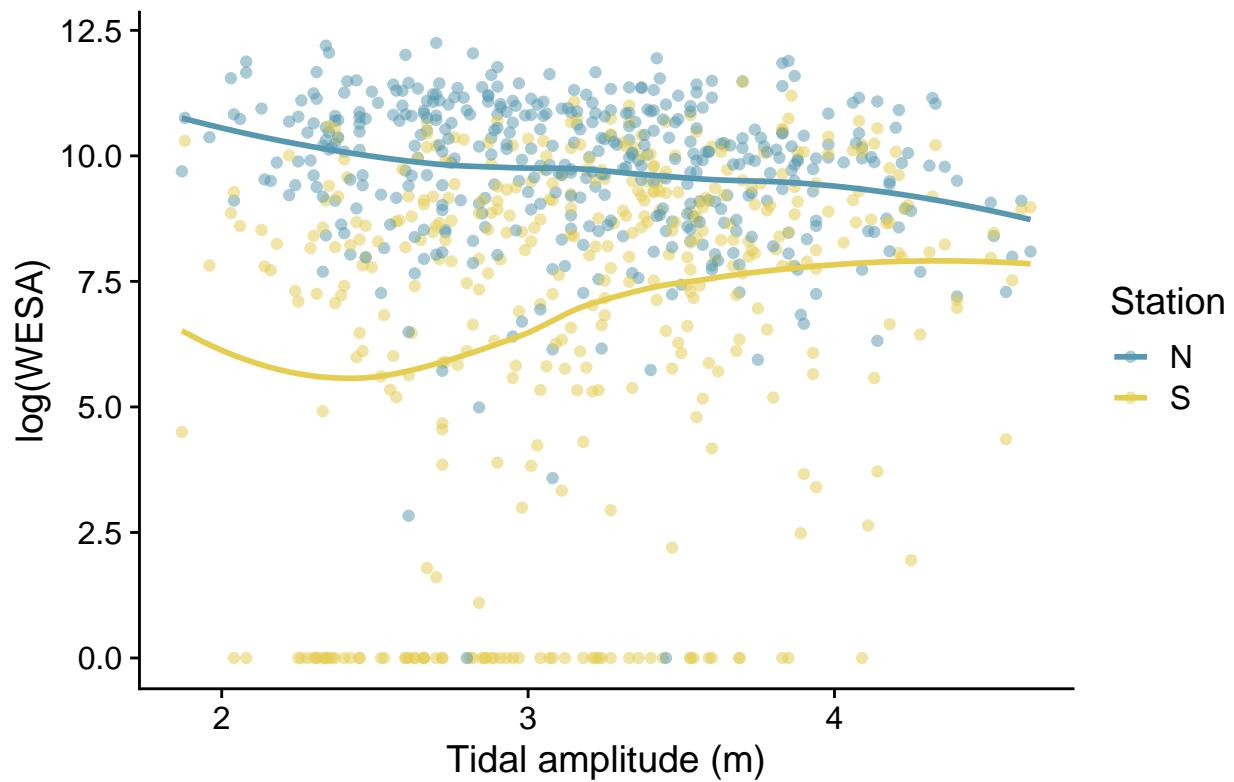
v (southerly winds) vs log(WESA)



Wind speed vs log(WESA)



Tidal amplitude vs log(WESA)



5.2 WESA model

5.2.1 Full model

```

## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
## Formula:
## "log_wesa ~ n_s*scale(flow) + n_s*scale(elev_range) + n_s*scale(u) + n_s*scale(v) + n_s*scale(windspd)
## Data: dat3
##
## REML criterion at convergence: 3645.9
##
## Scaled residuals:
##    Min     1Q Median     3Q    Max
## -4.2143 -0.2964  0.1061  0.5668  2.1699
##
## Random effects:
## Groups   Name        Variance Std.Dev. Corr
## year     (Intercept) 0.40916  0.6397
##           dos          0.56288  0.7503  -0.19
##           I(dos^2)     0.05842  0.2417  -0.28  0.72
## Residual            4.38790  2.0947
## Number of obs: 823, groups: year, 24
##
## Fixed effects:
##                               Estimate Std. Error      df t value Pr(>|t|)
## (Intercept)                10.930360  0.184098  25.276326 59.372 < 2e-16 ***
## n_sS                   -2.927783  0.146196 738.629067 -20.026 < 2e-16 ***
## scale(flow)              -0.231610  0.162326  89.064901 -1.427 0.15713
## scale(elev_range)         -0.105303  0.109812 662.018332 -0.959 0.33794
## scale(u)                  0.154439  0.128462 774.260380  1.202 0.22965
## scale(v)                  0.031740  0.139251 779.094862  0.228 0.81976
## scale(windspd)            -0.007841  0.116143 775.508353 -0.068 0.94619
## dos                      0.080552  0.198923  28.285788  0.405 0.68857
## I(dos^2)                 -1.368286  0.100038 15.304719 -13.678 5.45e-10 ***
## year_c                   0.184449  0.143744 19.459908  1.283 0.21450
## n_sS:scale(flow)          -0.084143  0.147250 738.058295 -0.571 0.56788
## n_sS:scale(elev_range)    1.119289  0.147184 738.188463  7.605 8.70e-14 ***
## n_sS:scale(u)             -0.466105  0.177009 737.799263 -2.633 0.00863 **
## n_sS:scale(v)             0.552910  0.191881 738.403236  2.882 0.00407 **
## n_sS:scale(windspd)       0.462816  0.161862 739.870872  2.859 0.00436 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

5.2.2 Backwards stepwise selection

```

## Backward reduced random-effect table:
##
##                                         Eliminated npar logLik AIC LRT Df
## <none>                                22 -1823.0 3689.9
## dos in (dos + I(dos^2) | year)          0   19 -1838.9 3715.8 31.822  3
## I(dos^2) in (dos + I(dos^2) | year)      0   19 -1824.8 3687.5 3.601  3
##                                         Pr(>Chisq)

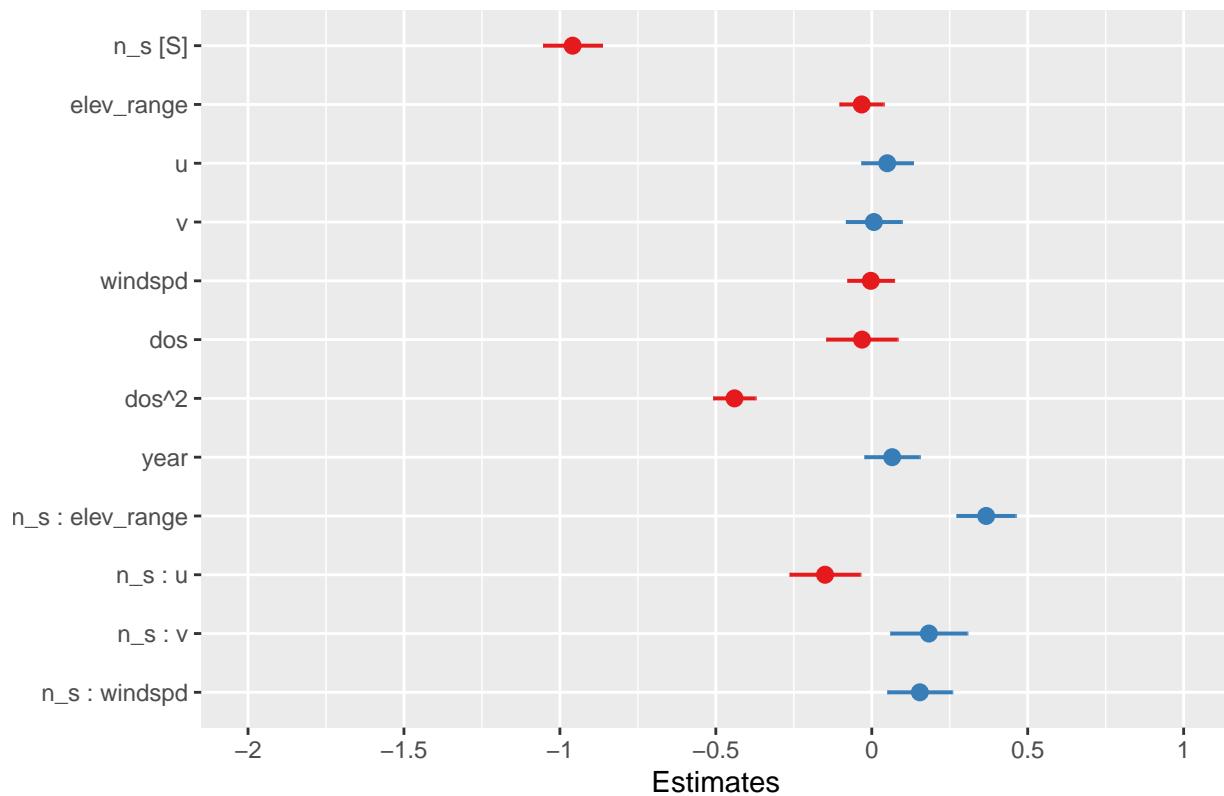
```

```

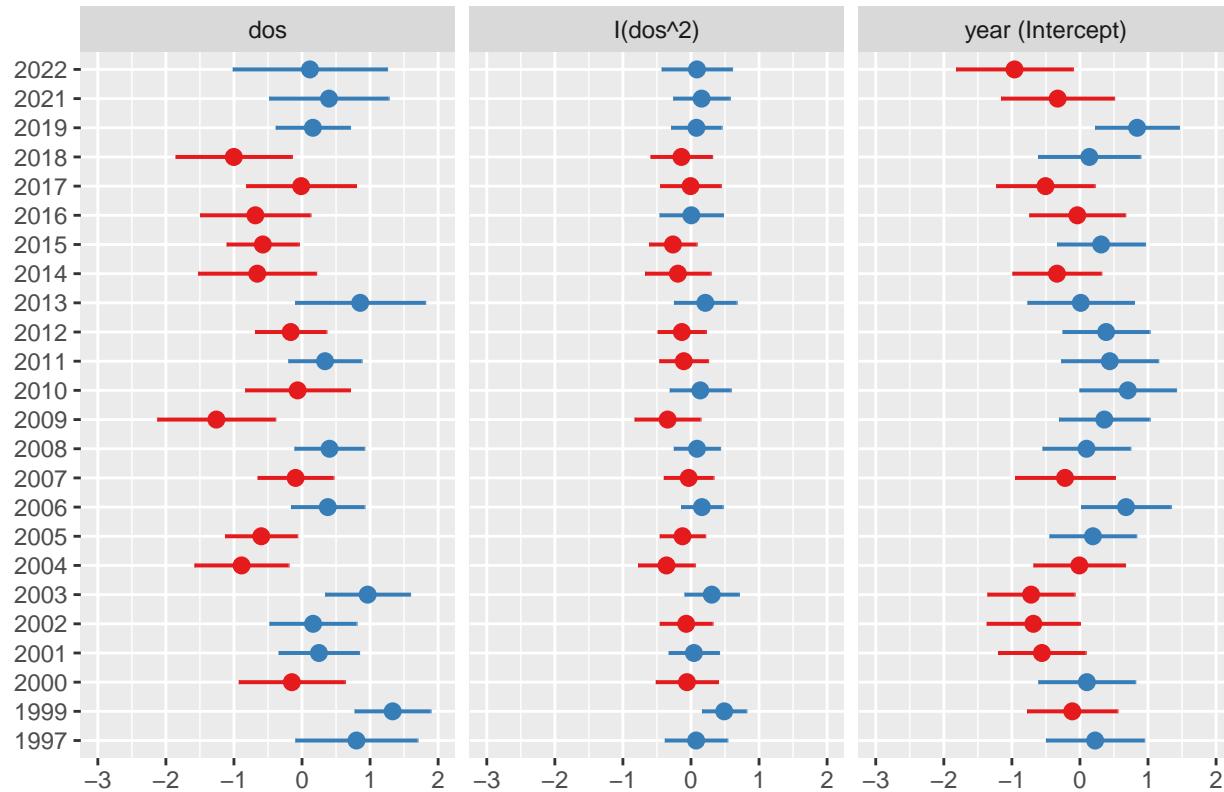
## <none>
## dos in (dos + I(dos^2) | year)      5.705e-07 ***
## I(dos^2) in (dos + I(dos^2) | year)   0.3079
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Backward reduced fixed-effect table:
## Degrees of freedom method: Satterthwaite
##
##          Eliminated Sum Sq Mean Sq NumDF DenDF F value    Pr(>F)
## n_s:scale(flow)           1   1.43   1.43     1 738.06  0.3265 0.567881
## scale(flow)              2  15.43  15.43     1  58.31  3.5200 0.065641
## dos                      0   1.32   1.32     1  19.60  0.3005 0.589751
## I(dos^2)                 0 710.84  710.84     1  15.56 161.7215 1.249e-09
## year_c                   0   9.24   9.24     1  21.40  2.1023 0.161573
## n_s:scale(elev_range)    0 253.55  253.55     1 738.71  57.6849 9.317e-14
## n_s:scale(u)              0  29.50   29.50     1 738.12  6.7111 0.009771
## n_s:scale(v)              0  37.36   37.36     1 738.82  8.4998 0.003659
## n_s:scale(windspd)        0  37.22   37.22     1 740.19  8.4682 0.003722
##
## n_s:scale(flow)
## scale(flow)             .
## dos
## I(dos^2)                 ***
## year_c
## n_s:scale(elev_range) ***
## n_s:scale(u)               **
## n_s:scale(v)               **
## n_s:scale(windspd)        **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Model found:
## log_wesa ~ n_s + scale(elev_range) + scale(u) + scale(v) + scale(windspd) + dos + I(dos^2) + year_c

```

Final WESA model – standardized fixed effect sizes

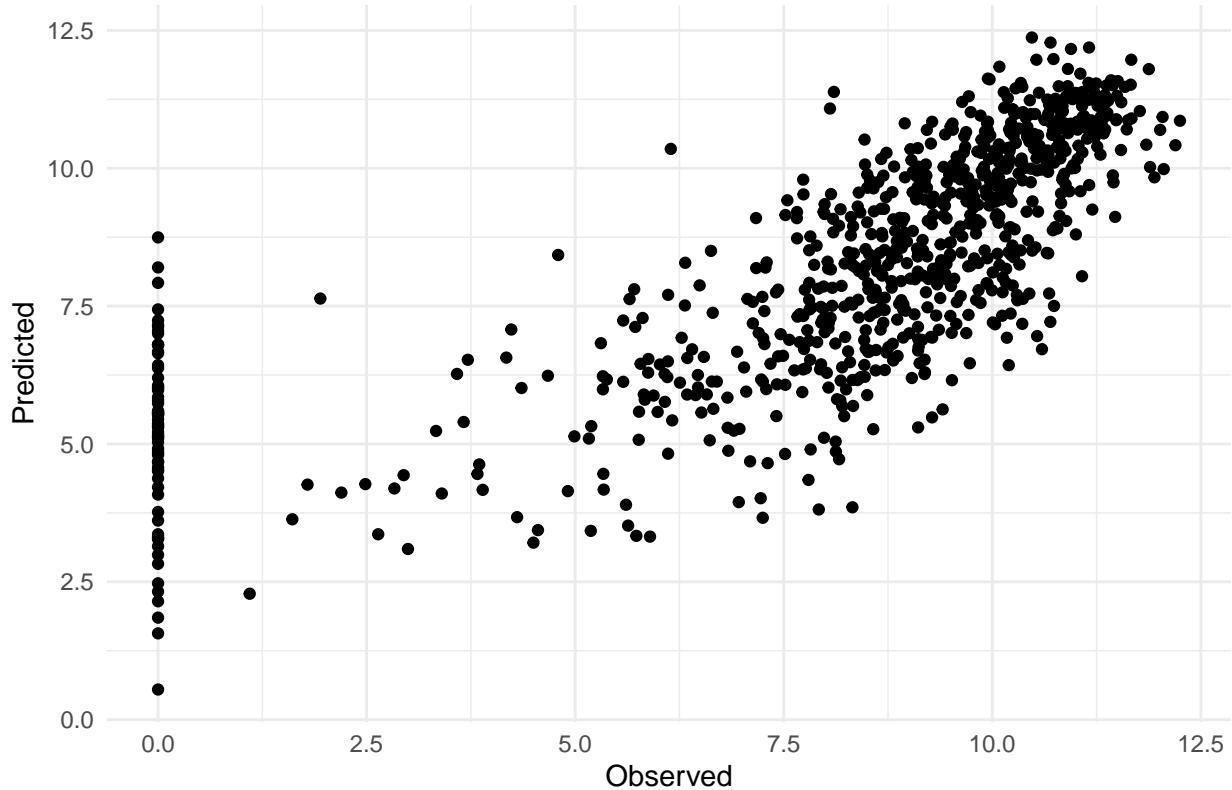


Random effects



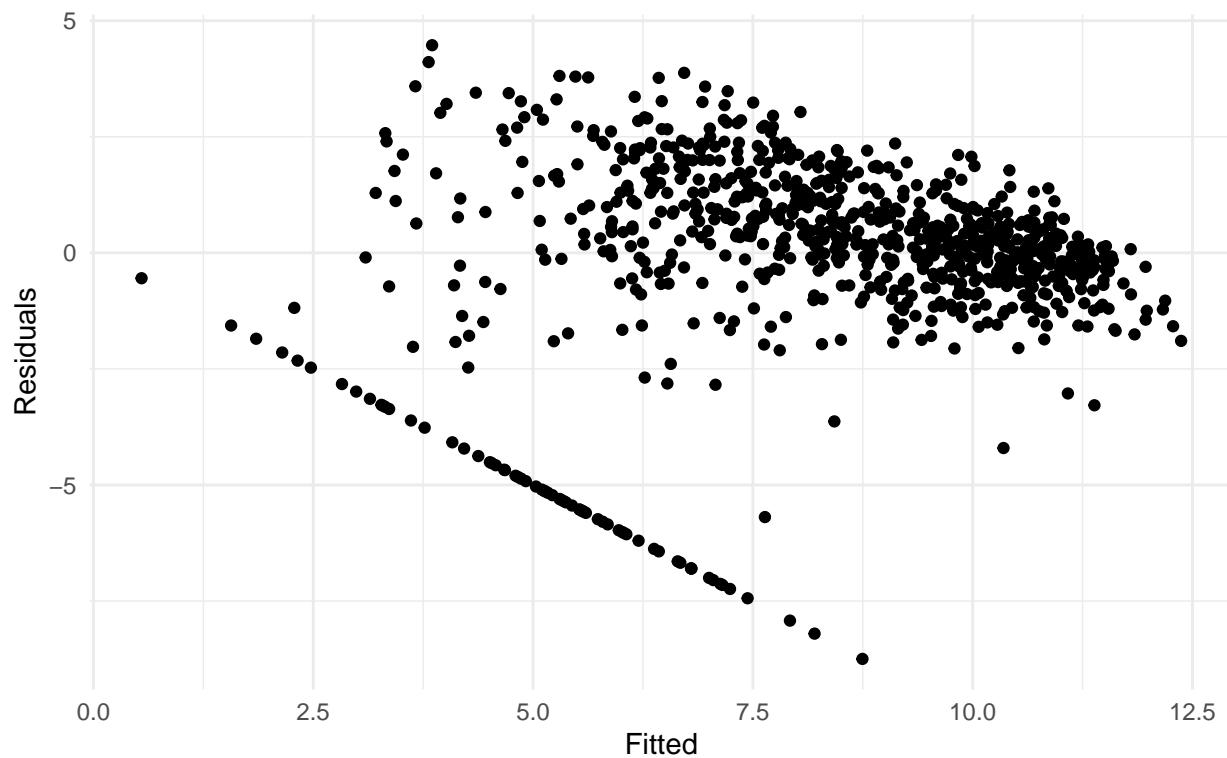
5.3 Final model diagnostics

Observed vs. Fitted values

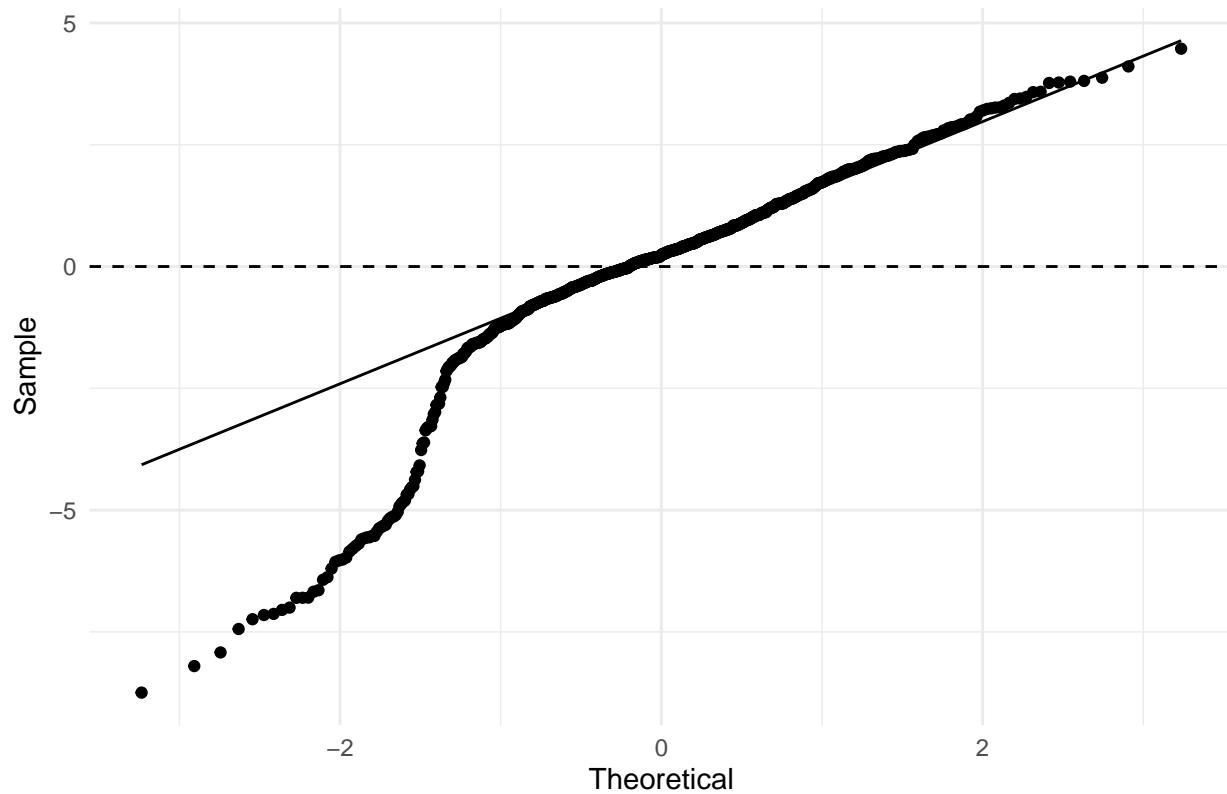


Heteroskedasticity

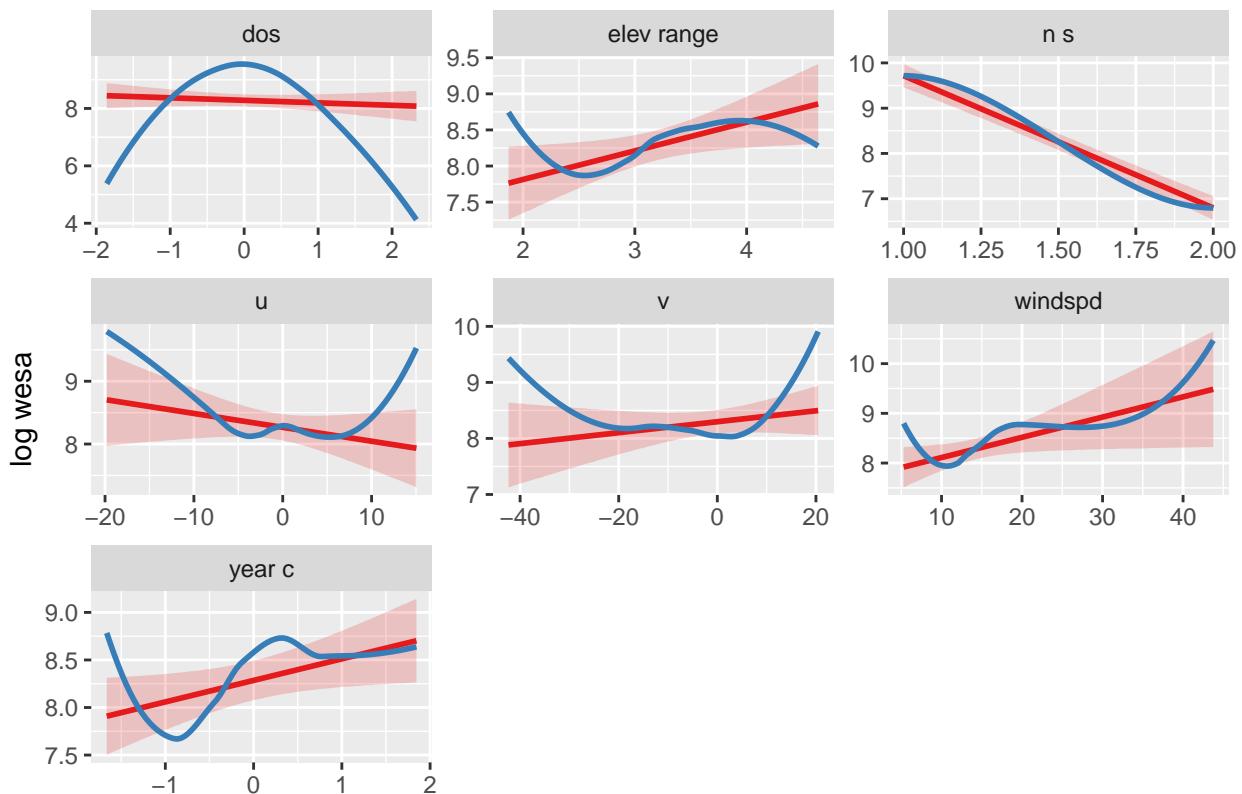
Fitted values vs. Residuals



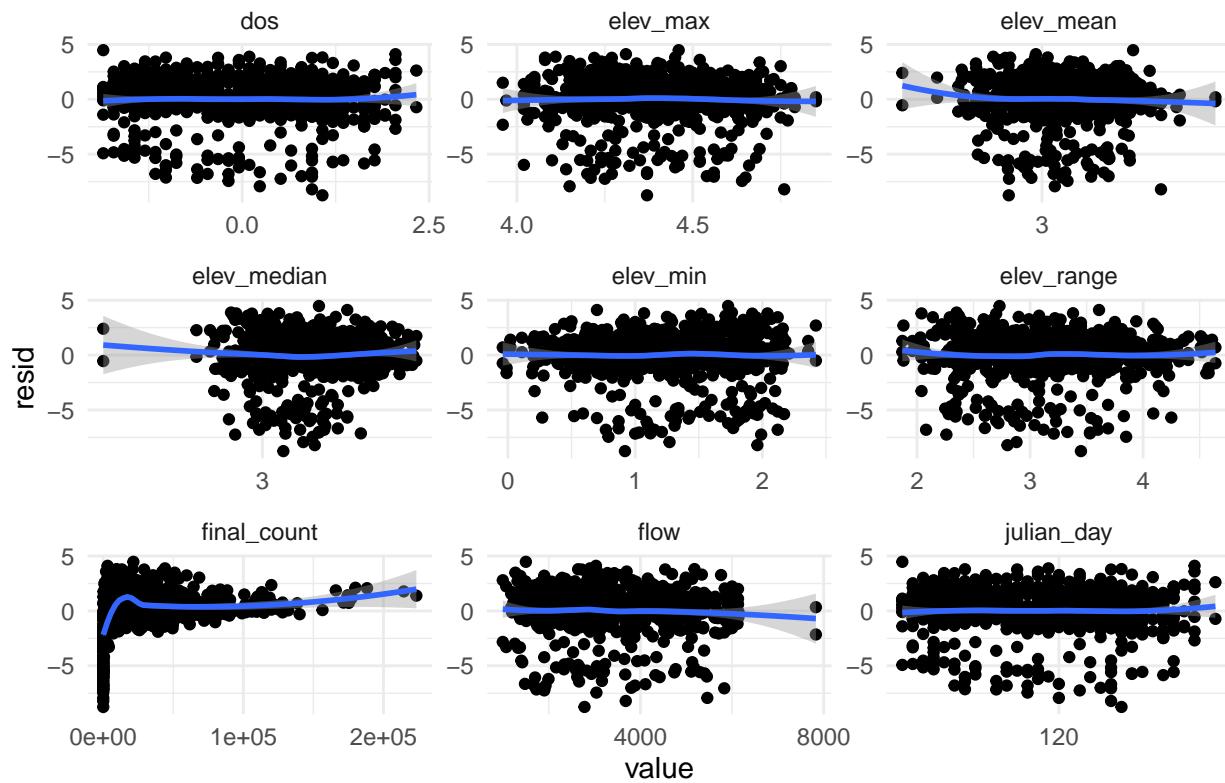
Quantile–Quantile

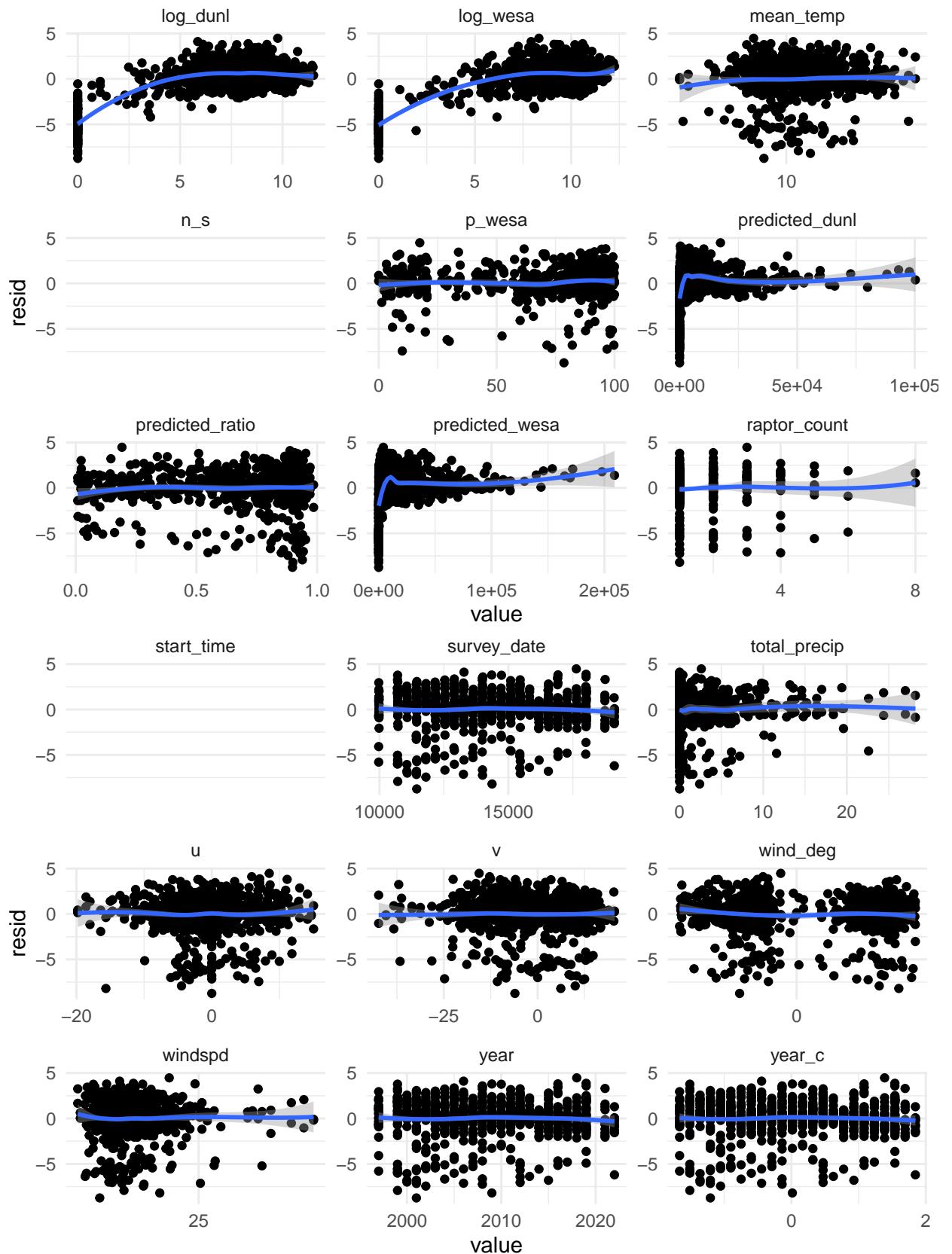


Coefficient slopes vs Response



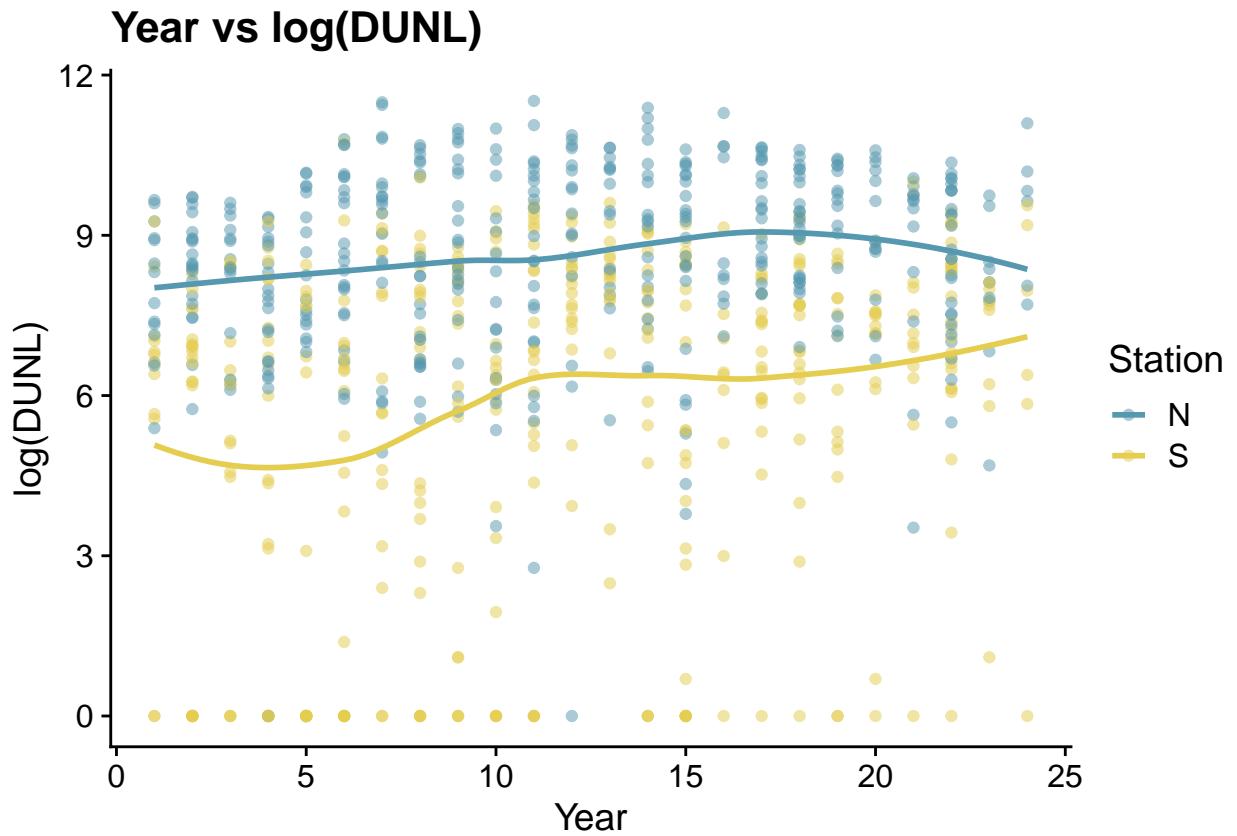
Full dataset variables vs. Residuals



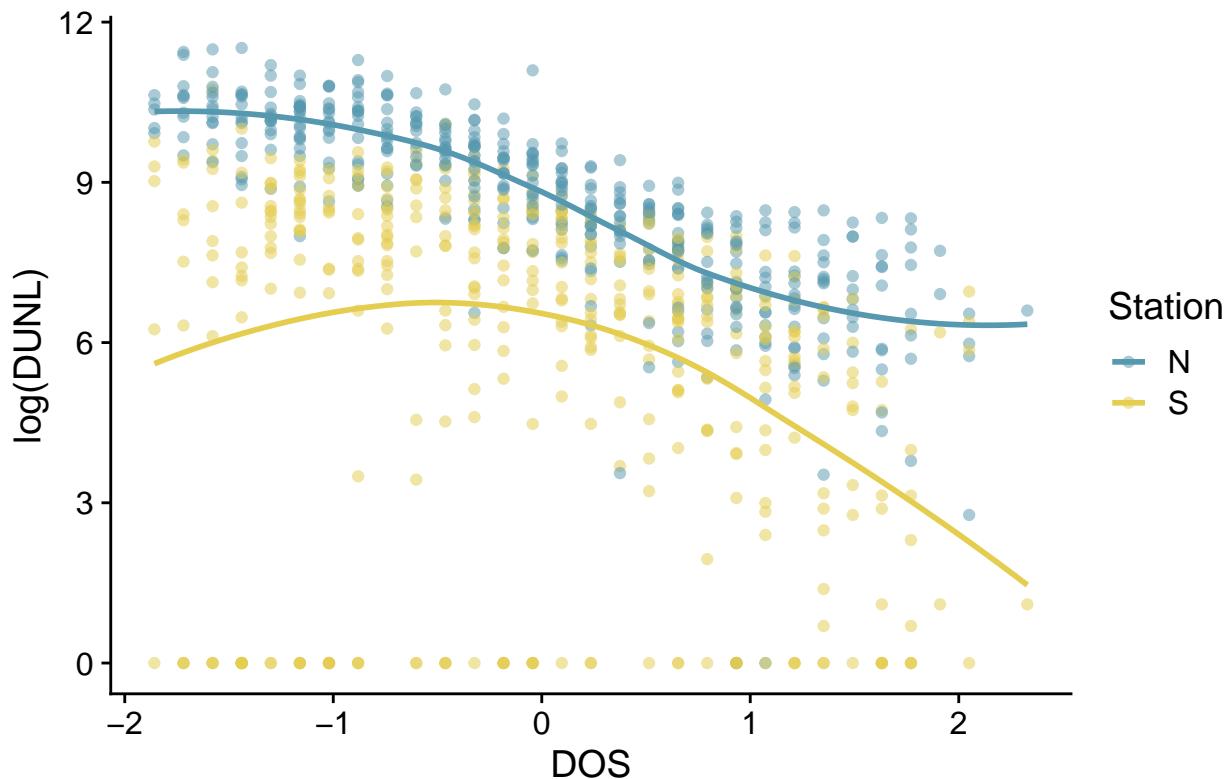


6 Final DUNL model

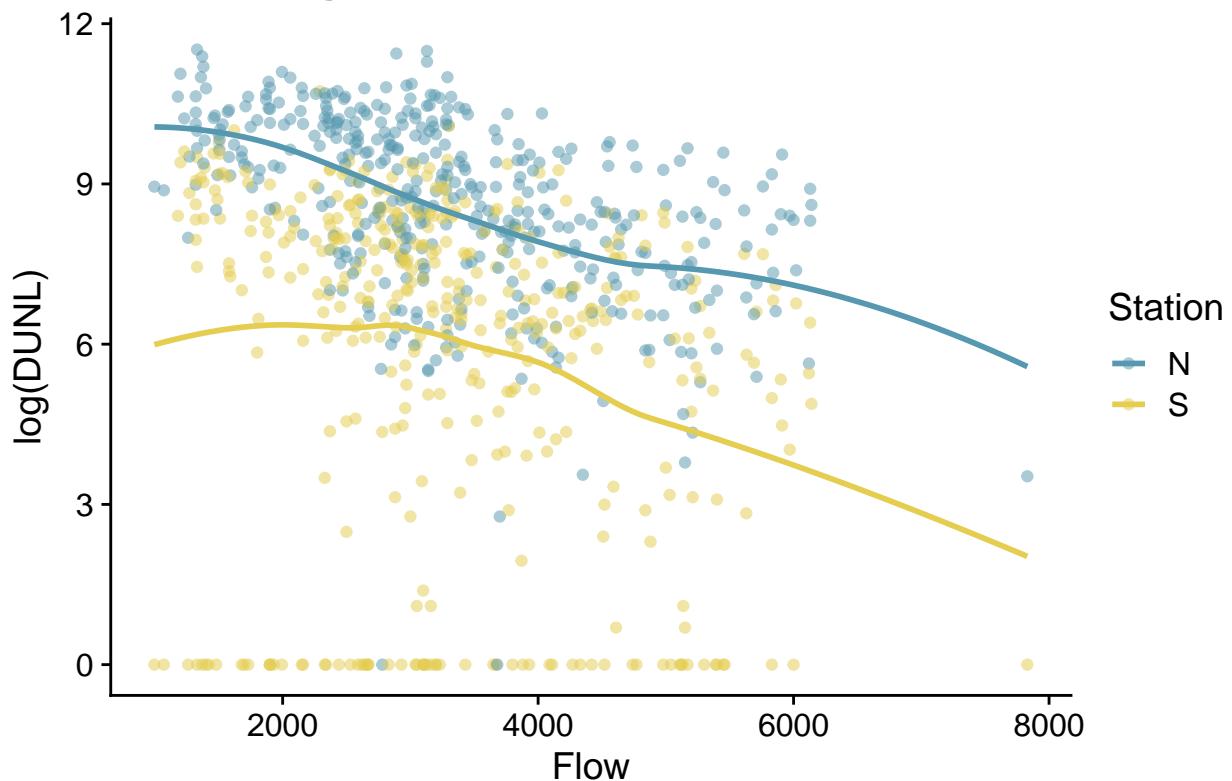
6.1 DUNL vs. variable plots



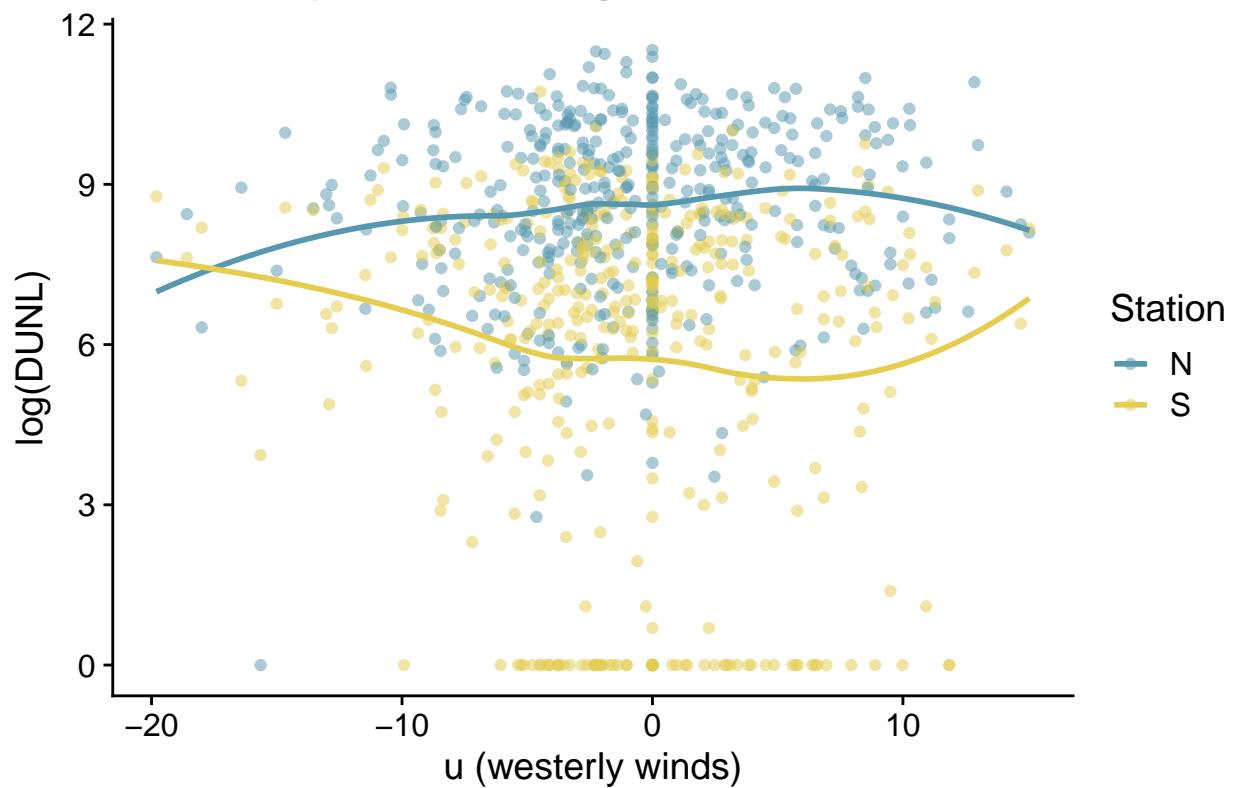
Day of Season vs log(DUNL)



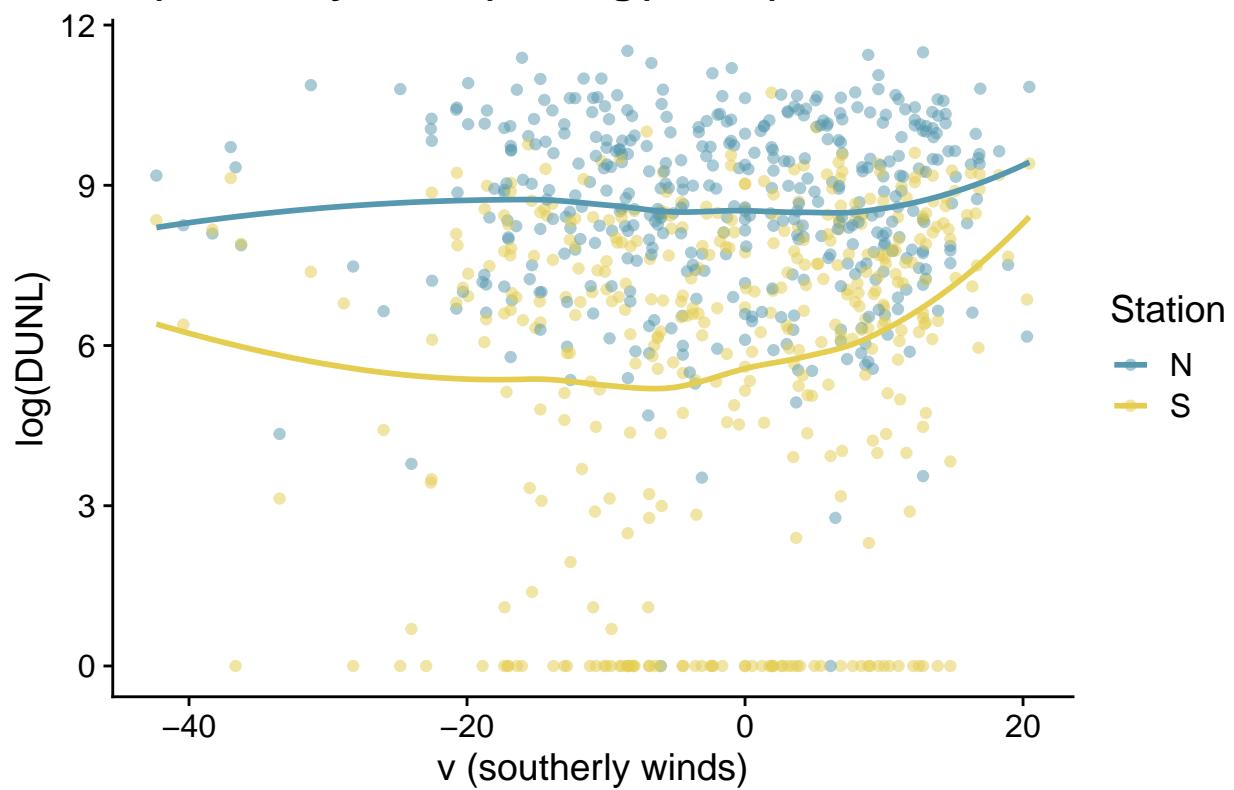
Flow vs log(DUNL)



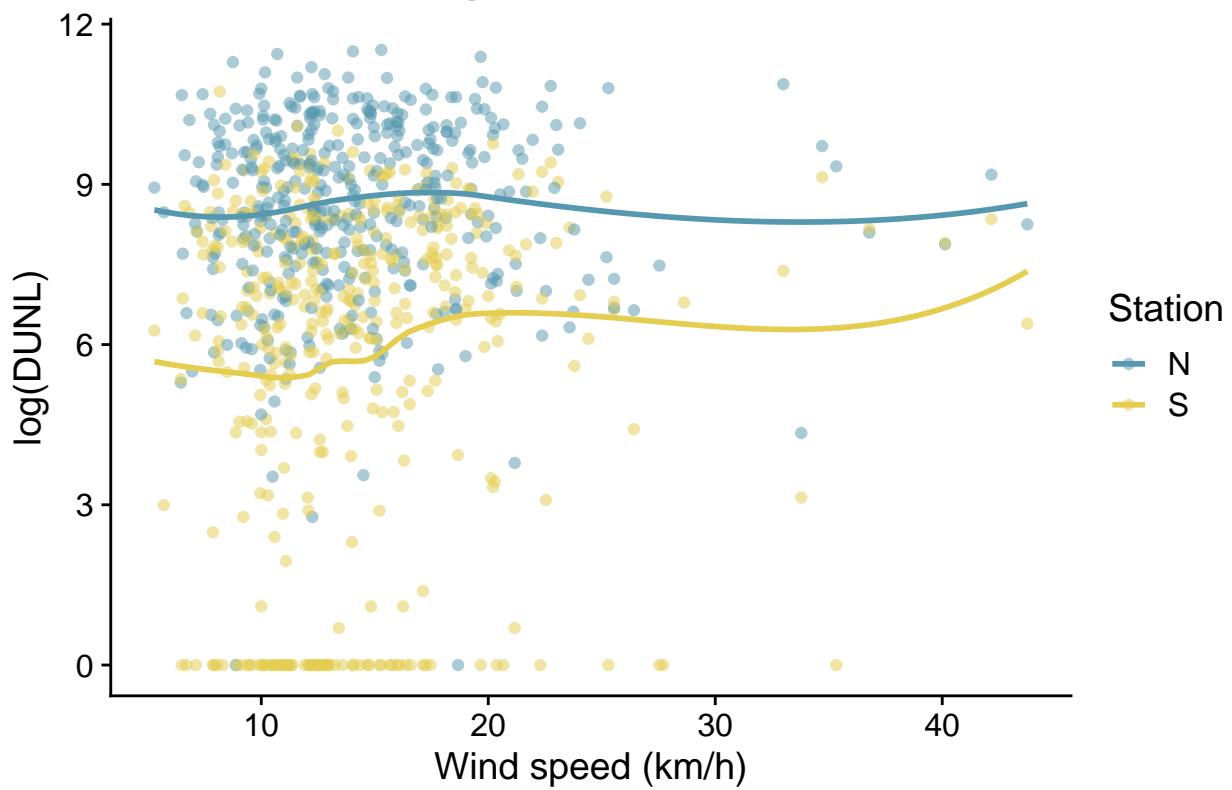
u (westerly winds) vs log(DUNL)



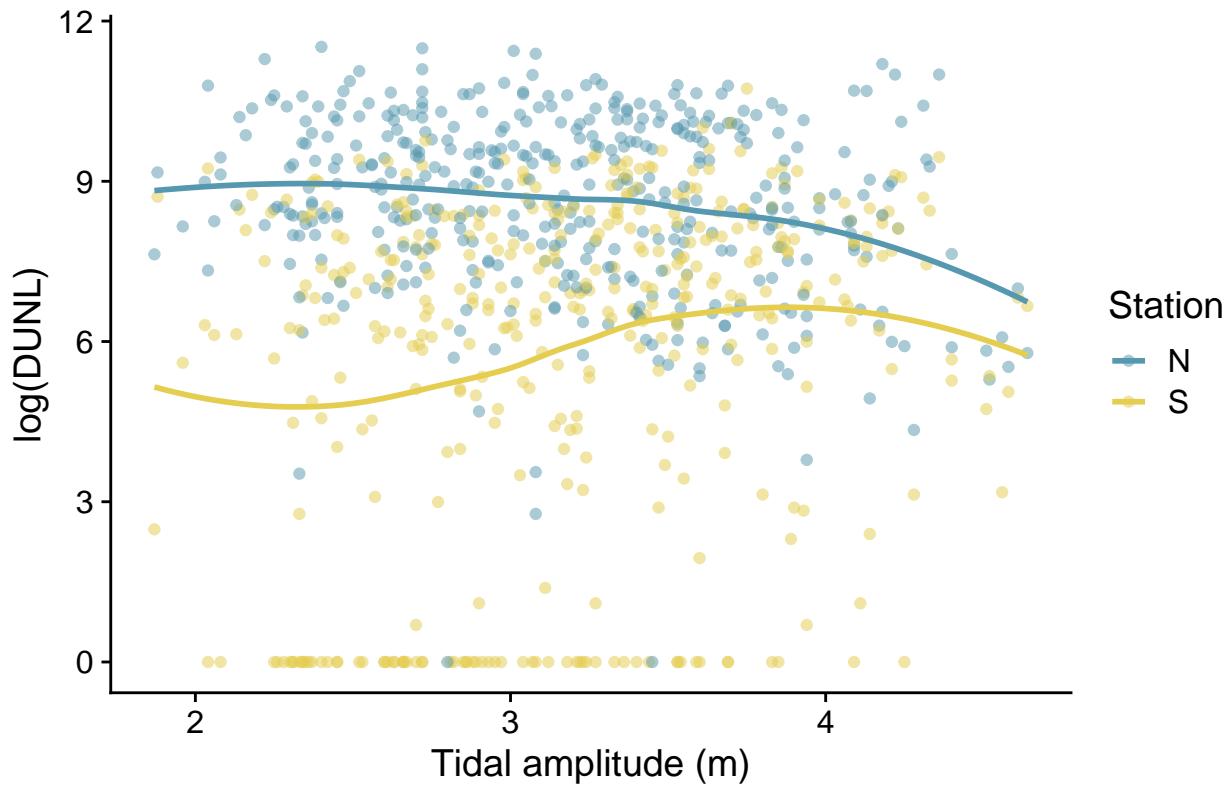
v (southerly winds) vs log(DUNL)



Wind speed vs log(DUNL)



Tidal amplitude vs log(DUNL)



6.2 DUNL model

6.2.1 Full model

```
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [  
## lmerModLmerTest]  
## Formula:  
## "log_dunl ~ n_s*scale(flow) + n_s*scale(elev_range) + n_s*scale(u) + n_s*scale(v) + n_s*scale(windspd)  
## Data: dat3  
##  
## REML criterion at convergence: 3503.4  
##  
## Scaled residuals:  
##      Min     1Q Median     3Q    Max  
## -3.7595 -0.3217  0.1066  0.5931  2.2032  
##  
## Random effects:  
## Groups   Name        Variance Std.Dev. Corr  
## year     (Intercept) 0.2260   0.4754  
##           dos         0.3355   0.5792  -0.27  
##           I(dos^2)    0.1507   0.3882  -0.61  0.02  
## Residual    3.7019   1.9240  
## Number of obs: 823, groups: year, 24  
##  
## Fixed effects:  
##                   Estimate Std. Error      df t value Pr(>|t|)  
## (Intercept)      9.08764   0.15393 32.69644 59.038 < 2e-16 ***  
## n_sS            -2.81293   0.13428 745.29435 -20.948 < 2e-16 ***  
## scale(flow)     -0.36901   0.13394 79.76843 -2.755 0.00727 **  
## scale(elev_range) -0.10177   0.10194 782.04887 -0.998 0.31843  
## scale(u)         0.08623   0.11838 792.08268  0.728 0.46658  
## scale(v)         -0.02025   0.12817 789.79617 -0.158 0.87451  
## scale(windspd)   -0.06444   0.10709 790.99022 -0.602 0.54754  
## dos              -0.86746   0.16245 27.76925 -5.340 1.12e-05 ***  
## I(dos^2)         -0.58502   0.11516 21.06177 -5.080 4.92e-05 ***  
## year_c          0.35374   0.10197 23.75414  3.469 0.00201 **  
## n_sS:scale(flow) 0.11244   0.13523 745.03029  0.831 0.40598  
## n_sS:scale(elev_range) 0.99966   0.13518 745.23157  7.395 3.81e-13 ***  
## n_sS:scale(u)    -0.35775   0.16259 744.25827 -2.200 0.02809 *  
## n_sS:scale(v)    0.53410   0.17627 744.67632  3.030 0.00253 **  
## n_sS:scale(windspd) 0.42252   0.14870 745.67061  2.841 0.00461 **  
## ---  
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

6.2.2 Backwards stepwise selection

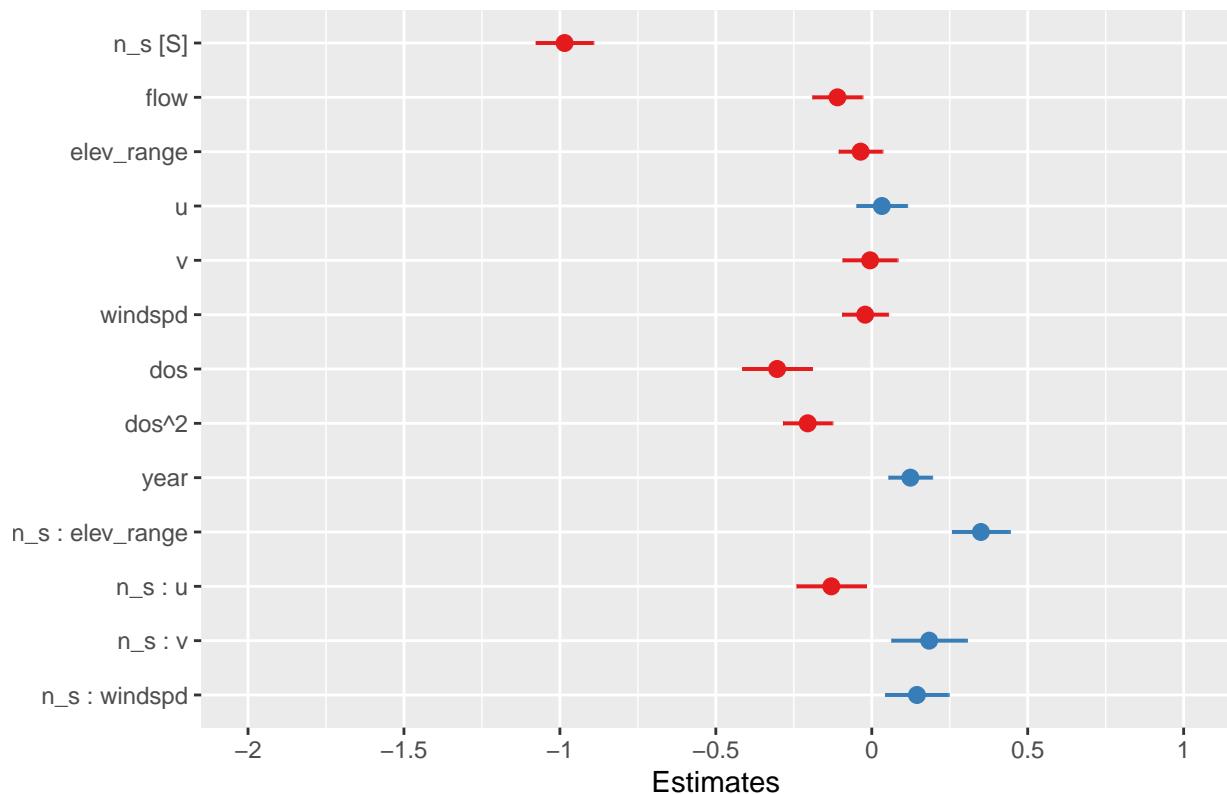
```
## Backward reduced random-effect table:  
##  
##                                         Eliminated npar logLik AIC LRT Df  
## <none>                                         22 -1751.7 3547.4  
## dos in (dos + I(dos^2) | year)             0   19 -1765.3 3568.7 27.303  3  
## I(dos^2) in (dos + I(dos^2) | year)         0   19 -1756.7 3551.4 10.015  3  
##                                         Pr(>Chisq)
```

```

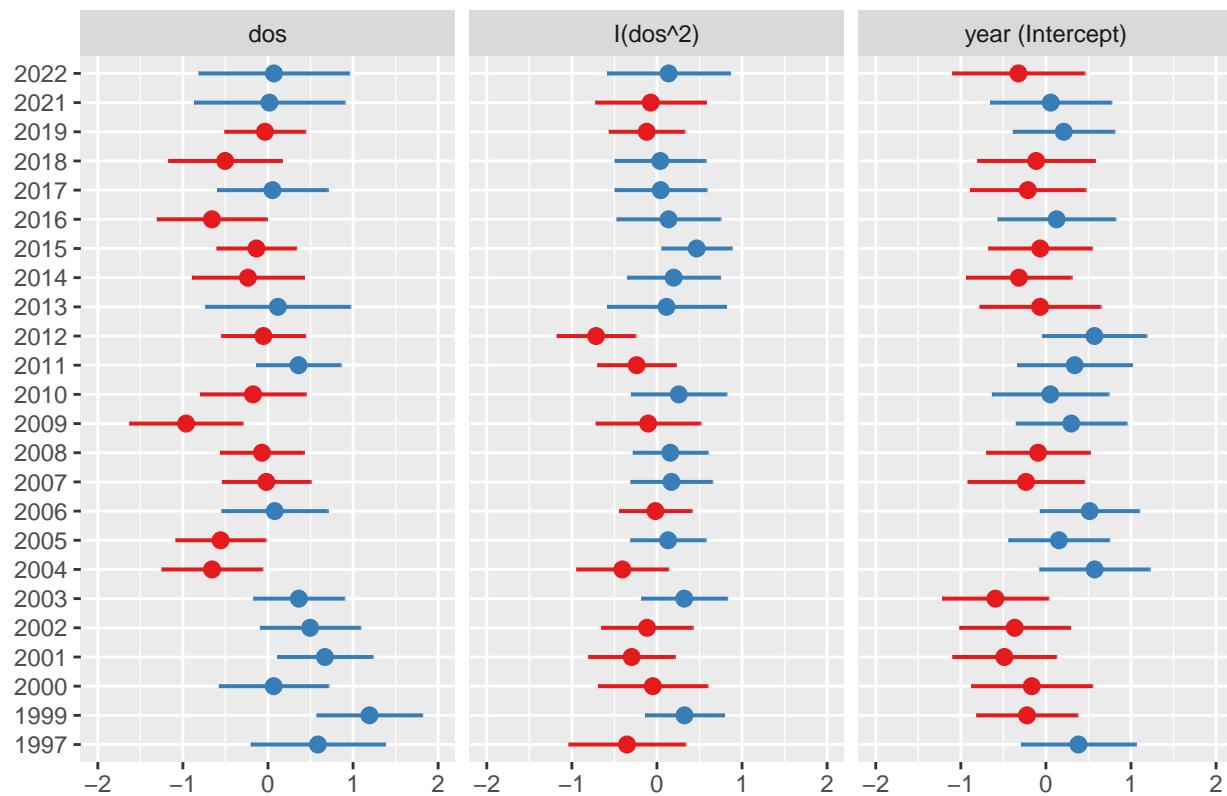
## <none>
## dos in (dos + I(dos^2) | year)      5.085e-06 ***
## I(dos^2) in (dos + I(dos^2) | year)  0.01844 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Backward reduced fixed-effect table:
## Degrees of freedom method: Satterthwaite
##
##          Eliminated   Sum Sq Mean Sq NumDF DenDF F value    Pr(>F)
## n_s:scale(flow)           1  2.559  2.559     1 745.03  0.6913  0.405978
## scale(flow)              0 26.903 26.903     1  46.75  7.2703  0.009715
## dos                      0 105.481 105.481     1  27.79 28.5050 1.123e-05
## I(dos^2)                 0  95.648  95.648     1  21.05 25.8477 4.879e-05
## year_c                   0  44.722  44.722     1  23.75 12.0855  0.001974
## n_s:scale(elev_range)    0 202.397 202.397     1 746.23 54.6954 3.791e-13
## n_s:scale(u)              0  19.401  19.401     1 745.19  5.2429  0.022316
## n_s:scale(v)              0  33.047  33.047     1 745.67  8.9305  0.002897
## n_s:scale(windspd)        0  28.803  28.803     1 746.67  7.7836  0.005407
##
## n_s:scale(flow)
## scale(flow)             **
## dos                     ***
## I(dos^2)                ***
## year_c                  **
## n_s:scale(elev_range)   ***
## n_s:scale(u)              *
## n_s:scale(v)              **
## n_s:scale(windspd)        **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Model found:
## log_dunl ~ n_s + scale(flow) + scale(elev_range) + scale(u) + scale(v) + scale(windspd) + dos + I(dos^2)

```

Final DUNL model – standardized fixed effect sizes

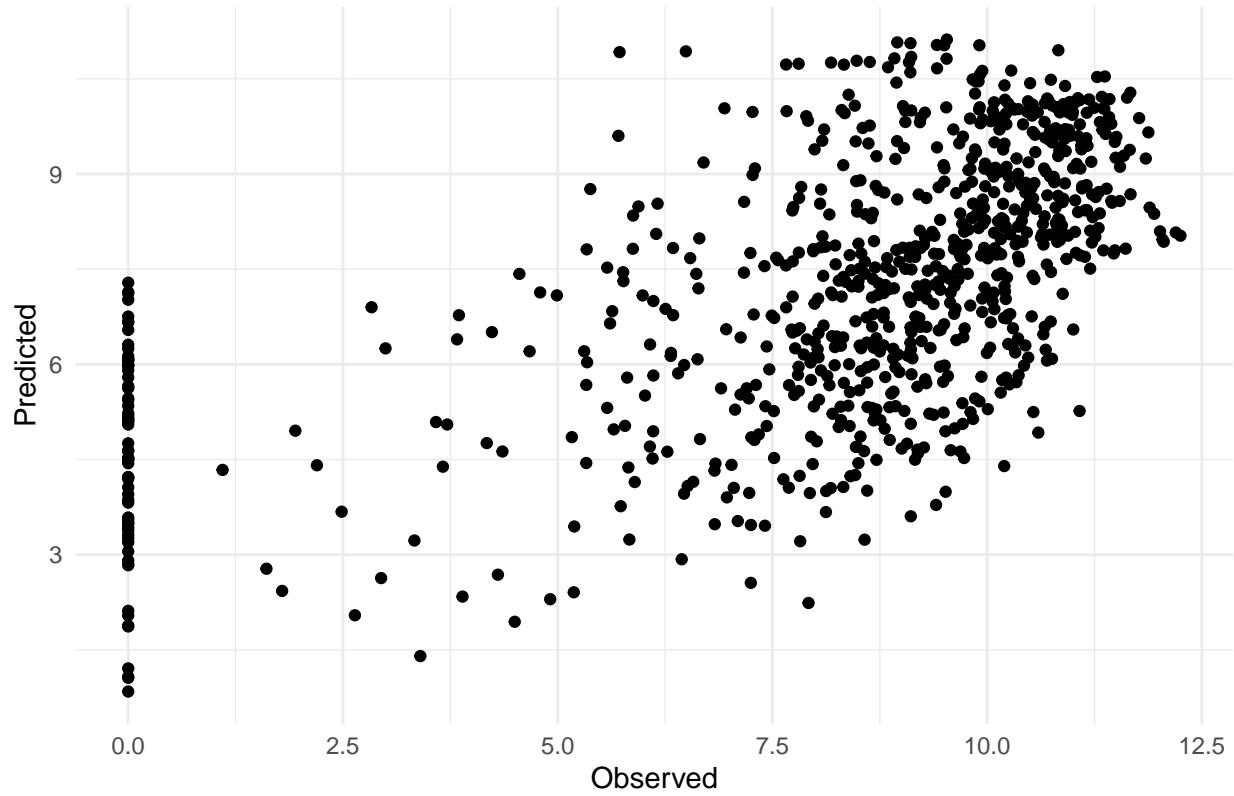


Random effects

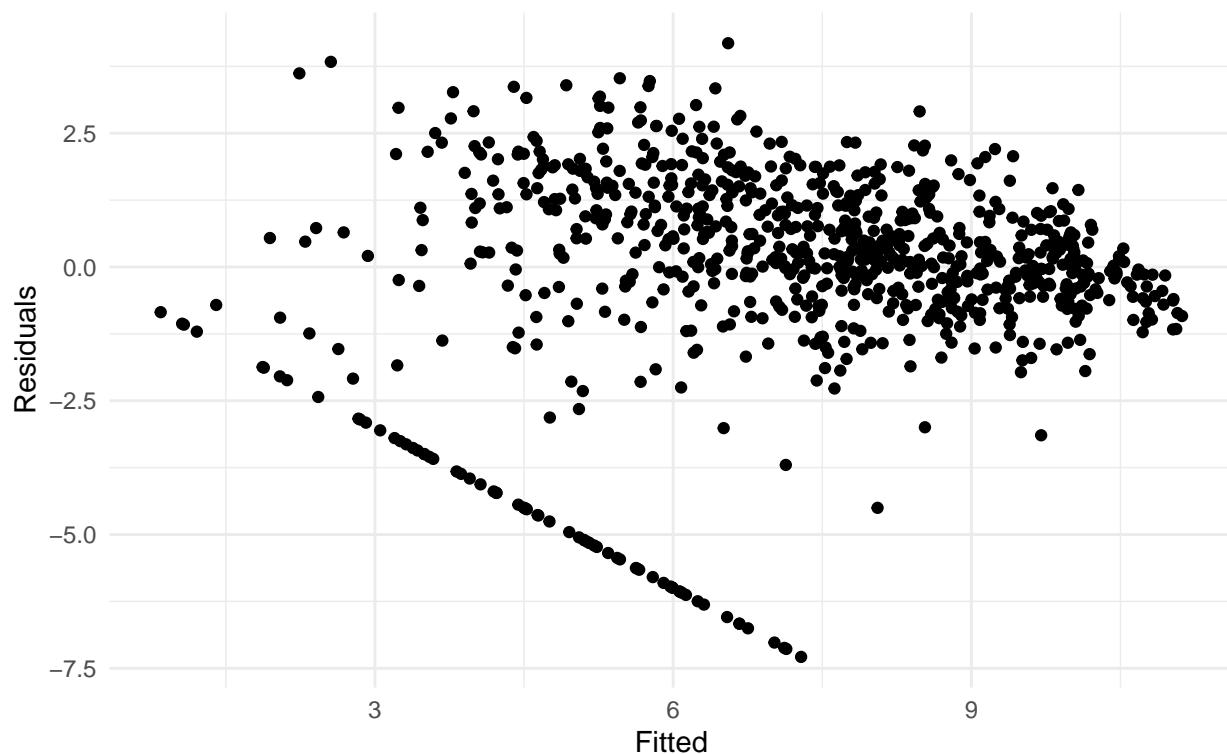


6.3 Final model diagnostics

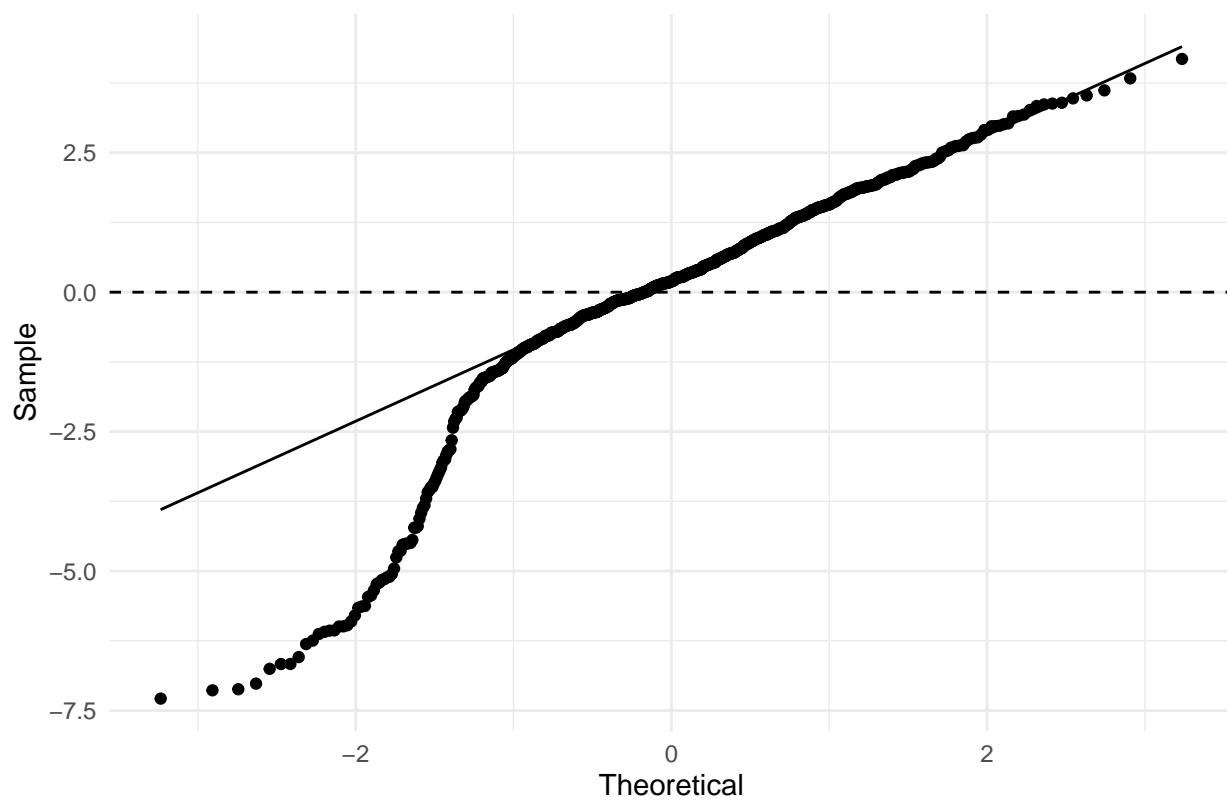
Observed vs. Fitted values



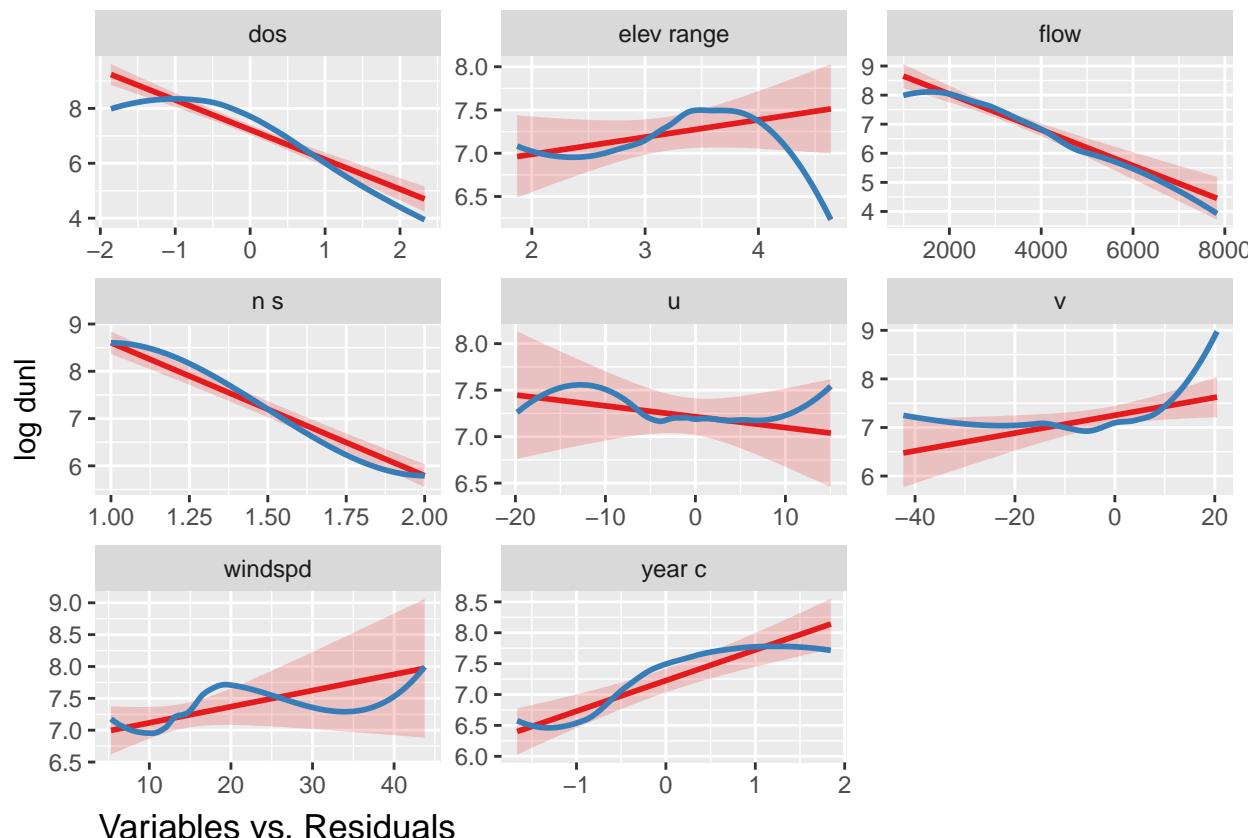
Heteroskedasticity Fitted values vs. Residuals



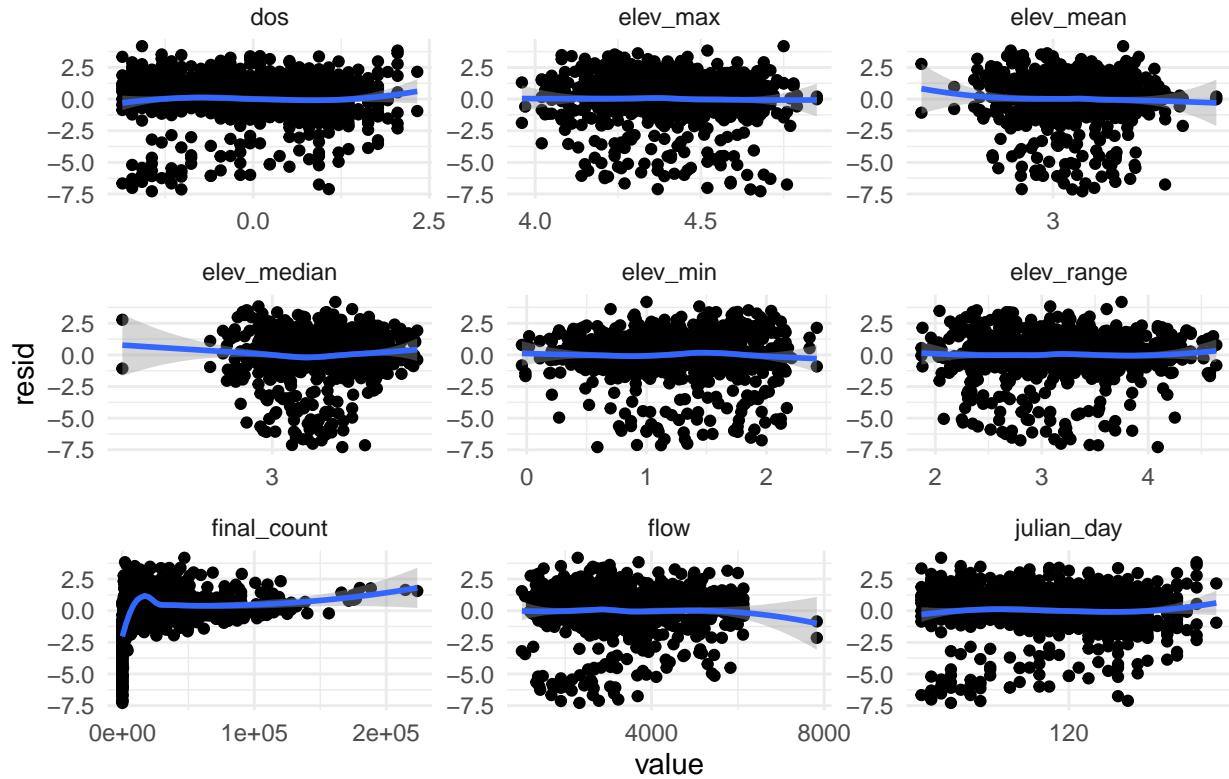
Quantile–Quantile

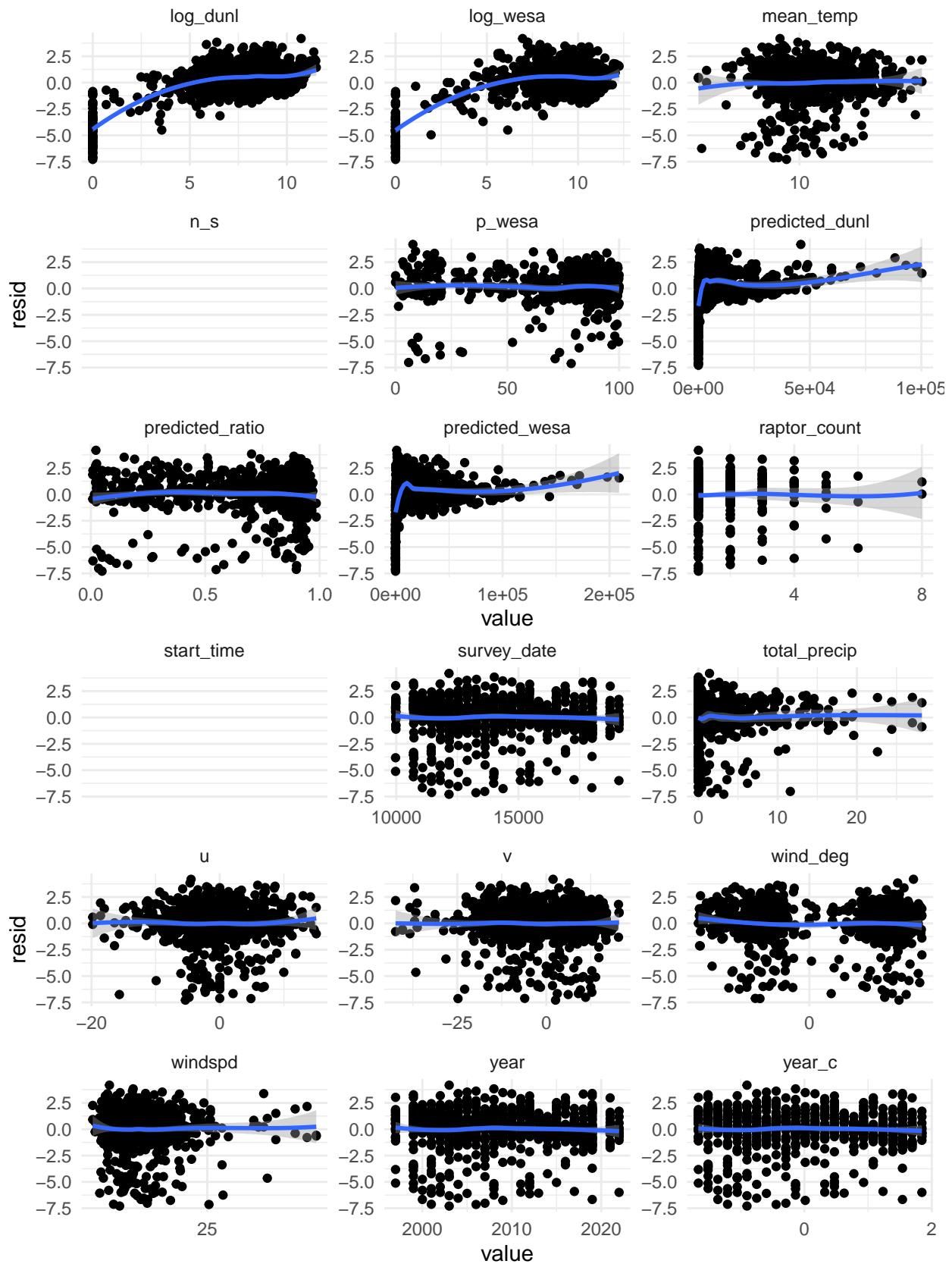


Coefficient slopes vs Response



Variables vs. Residuals





7 Yearly trends in peep population

The original Canham et al. (2021) paper provided yearly population trends of WESA and DUNL at Brunswick Point. These same population trends are extended here with 2020-2022 data, using the same population models derived in the orginal paper (i.e., no north vs. south component). Values represent population indices (with 95% confidence intervals) calculated as predicted values for each year from final models for each species, with independent variables held at median values for each year. Daily total counts (not broken down by survey station) are used as the data for this model; differences between the original Canham data is that this dataset only goes back to 1994 (rather than 1991) and goes up to 2022 (rather than 2019).

WESA model:

$$\log(WESA) \sim \text{year} + \text{dos} + I(\text{dos}^2) + \text{scale}(\text{elev.range}) + \text{scale}(\text{flow}) + \text{scale}(u) + (\text{dos} + I(\text{dos}^2) | \text{year})$$

DUNL model:

$$\log(DUNL) \sim \text{year} + \text{dos} + I(\text{dos}^2) + \text{scale}(\text{elev.range}) + \text{scale}(\text{flow}) + \text{scale}(u) + (\text{dos} + I(\text{dos}^2) | \text{year})$$

7.1 Yearly population trend model summaries

The westerly wind vector, u, is not significant in either model, deviating from the original Canham paper. As in the original Canham paper, however, `year` remains insignificant in the DUNL model.

7.1.1 WESA

```
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [  
## lmerModLmerTest]  
## Formula:  
## log_wesa ~ year_c + dos + I(dos^2) + scale(elev_range) + scale(flow) +  
## scale(u) + (dos + I(dos^2) | year)  
## Data: dt  
## Control: lme4::lmerControl(optimizer = "bobyqa")  
##  
## REML criterion at convergence: 1256.5  
##  
## Scaled residuals:  
##      Min       1Q   Median       3Q      Max  
## -5.6828 -0.3770  0.0445  0.4588  3.0829  
##  
## Random effects:  
## Groups   Name        Variance Std.Dev. Corr  
## year     (Intercept) 0.3393  0.5825  
##           dos         0.4120  0.6418   0.21  
##           I(dos^2)   0.3483  0.5901  -0.33  0.15  
## Residual    0.5510  0.7423  
## Number of obs: 474, groups: year, 26  
##  
## Fixed effects:  
##             Estimate Std. Error      df t value Pr(>|t|)  
## (Intercept) 11.25075  0.12571 18.27449 89.498 < 2e-16 ***  
## year_c     -0.28230  0.10505 21.12458 -2.687  0.01375 *  
## dos        -0.52333  0.14966 30.99909 -3.497  0.00145 **  
## I(dos^2)   -1.64309  0.12550 19.60208 -13.093 3.81e-11 ***  
## scale(elev_range) -0.10934  0.04025 427.05983 -2.716  0.00687 **
```

```

## scale(flow)      -0.17804   0.09479 113.24428 -1.878  0.06292 .
## scale(u)        0.05764   0.03774 411.61758  1.527  0.12748
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##          (Intr) year_c dos   I(d^2) scl(_) scl(f)
## year_c    -0.061
## dos       0.204 -0.038
## I(dos^2) -0.382 -0.001  0.106
## scl(lv_rng) -0.018 -0.005 -0.060  0.008
## scale(flow) -0.089  0.072 -0.431  0.043  0.066
## scale(u)    -0.011 -0.006  0.005  0.007 -0.090  0.064

```

7.1.2 DUNL

```

## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
## Formula:
## log_dunl ~ year_c + dos + I(dos^2) + scale(elev_range) + scale(flow) +
## scale(u) + (dos + I(dos^2) | year)
## Data: dt
## Control: lme4::lmerControl(optimizer = "bobyqa")
##
## REML criterion at convergence: 1225.4
##
## Scaled residuals:
##    Min     1Q Median     3Q    Max
## -6.0878 -0.4109  0.0403  0.4687  2.8480
##
## Random effects:
## Groups   Name        Variance Std.Dev. Corr
## year     (Intercept) 0.3737   0.6113
##           dos         0.4504   0.6711  -0.04
##           I(dos^2)   0.4265   0.6531  -0.66  0.25
## Residual            0.5115   0.7152
## Number of obs: 474, groups: year, 26
##
## Fixed effects:
##             Estimate Std. Error    df t value Pr(>|t|)
## (Intercept) 9.21816   0.13010 21.48622 70.854 < 2e-16 ***
## year_c     -0.09842   0.09158 23.30418 -1.075  0.29351
## dos        -1.45448   0.15299 30.96880 -9.507 1.07e-10 ***
## I(dos^2)   -0.71386   0.13645 23.46328 -5.232 2.48e-05 ***
## scale(elev_range) -0.11852   0.03902 426.10854 -3.037 0.00254 **
## scale(flow)  -0.22829   0.08945 105.32789 -2.552 0.01214 *
## scale(u)     0.04964   0.03637 416.86452   1.365 0.17302
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##          (Intr) year_c dos   I(d^2) scl(_) scl(f)
## year_c    -0.050

```

```
## dos          0.002 -0.037
## I(dos^2)    -0.653 -0.004  0.194
## scl(lv_rng) -0.019  0.010 -0.063  0.007
## scale(flow) -0.080  0.059 -0.405  0.038  0.077
## scale(u)     -0.010 -0.008  0.008  0.006 -0.089  0.057
```

