

# Seasonal Difference Exploration

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## Explore the seasonal differences

```
load("./dt_long.RData")
load("./ID_in.RData")
load("./beta.res.postmean.RData")

dt_season <-
  dt_long %>%
  drop_na() %>%
  filter(ID %in% ID_in) %>%
  distinct(ID, .keep_all = TRUE) %>%
  select(ID, Season, Month, Nature) %>%
  mutate(Month = factor(Month, levels = month.name))

season_diff <-
  merge(dt_season, beta.res.postmean, by = c("ID")) %>%
  janitor::clean_names()

# Beta0
intercept.fit <- lm(intercept ~ month + season + nature, data = season_diff)
# Beta1
wind_prev.fit <- lm(wind_prev ~ month + season + nature, data = season_diff)
# Beta2
lat_change.fit <- lm(lat_change ~ month + season + nature, data = season_diff)
# Beta3
long_change.fit <- lm(long_change ~ month + season + nature, data = season_diff)
#Beta4
wind_change.fit <- lm(wind_change ~ month + season + nature, data = season_diff)

summary(intercept.fit)

##
## Call:
## lm(formula = intercept ~ month + season + nature, data = season_diff)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.43433 -0.03822  0.00234  0.04052  0.38709
##
## Coefficients:
```

```
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept)  4.4810021  0.3902677  11.482  <2e-16 ***
## monthApril   0.0232609  0.1113880   0.209  0.8346
## monthMay     0.0259813  0.0942017   0.276  0.7828
## monthJune    0.0275693  0.0922175   0.299  0.7651
## monthJuly    0.0125400  0.0918533   0.137  0.8914
## monthAugust -0.0198034  0.0913669  -0.217  0.8285
## monthSeptember -0.0070528  0.0912856  -0.077  0.9384
## monthOctober  0.0093435  0.0913761   0.102  0.9186
## monthNovember 0.0145692  0.0924416   0.158  0.8748
## monthDecember 0.0057977  0.0976110   0.059  0.9527
## season      -0.0003419  0.0001895  -1.804  0.0717 .
## natureET     0.0008449  0.0298315   0.028  0.9774
## natureNR     0.0008122  0.0484835   0.017  0.9866
## natureSS     0.0141564  0.0205164   0.690  0.4904
## natureTS     0.0118370  0.0166932   0.709  0.4785
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.09022 on 682 degrees of freedom
## Multiple R-squared:  0.03095,    Adjusted R-squared:  0.01105
## F-statistic: 1.556 on 14 and 682 DF,  p-value: 0.0866
```

```
summary(wind_prev.fit)
```

```
##
## Call:
## lm(formula = wind_prev ~ month + season + nature, data = season_diff)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.135790 -0.019433  0.002041  0.022272  0.061680
##
## Coefficients:
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept)  1.343e+00  1.215e-01  11.058  < 2e-16 ***
## monthApril   1.479e-02  3.467e-02   0.427  0.669679
## monthMay     -1.180e-04  2.932e-02  -0.004  0.996789
## monthJune    5.393e-03  2.870e-02   0.188  0.850987
## monthJuly    1.540e-02  2.859e-02   0.539  0.590174
## monthAugust  2.332e-02  2.843e-02   0.820  0.412418
## monthSeptember 2.610e-02  2.841e-02   0.919  0.358560
## monthOctober 2.108e-02  2.844e-02   0.741  0.458718
## monthNovember 2.461e-02  2.877e-02   0.856  0.392526
## monthDecember 8.824e-03  3.038e-02   0.290  0.771531
## season      -2.252e-04  5.899e-05  -3.817  0.000147 ***
## natureET     3.733e-03  9.284e-03   0.402  0.687709
## natureNR     -1.461e-02  1.509e-02  -0.969  0.333111
## natureSS     -3.330e-03  6.385e-03  -0.522  0.602172
## natureTS     -5.998e-03  5.195e-03  -1.155  0.248693
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.02808 on 682 degrees of freedom
```

```
## Multiple R-squared:  0.0659, Adjusted R-squared:  0.04672
## F-statistic: 3.437 on 14 and 682 DF,  p-value: 1.998e-05
```

```
summary(lat_change.fit)
```

```
##
## Call:
## lm(formula = lat_change ~ month + season + nature, data = season_diff)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.90321 -0.07062  0.00781  0.07691  0.95935
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   4.131e-02  6.667e-01   0.062   0.951
## monthApril    1.656e-02  1.903e-01   0.087   0.931
## monthMay      7.088e-02  1.609e-01   0.440   0.660
## monthJune     -7.088e-03  1.575e-01  -0.045   0.964
## monthJuly     -9.091e-03  1.569e-01  -0.058   0.954
## monthAugust   -5.225e-02  1.561e-01  -0.335   0.738
## monthSeptember -3.611e-02  1.559e-01  -0.232   0.817
## monthOctober  -2.862e-02  1.561e-01  -0.183   0.855
## monthNovember  2.400e-02  1.579e-01   0.152   0.879
## monthDecember -5.431e-02  1.668e-01  -0.326   0.745
## season        3.655e-05  3.238e-04   0.113   0.910
## natureET      -7.020e-02  5.096e-02  -1.378   0.169
## natureNR       5.897e-03  8.283e-02   0.071   0.943
## natureSS      -1.352e-03  3.505e-02  -0.039   0.969
## natureTS      -1.545e-02  2.852e-02  -0.542   0.588
##
## Residual standard error: 0.1541 on 682 degrees of freedom
## Multiple R-squared:  0.02561,    Adjusted R-squared:  0.005609
## F-statistic:  1.28 on 14 and 682 DF,  p-value: 0.2137
```

```
summary(long_change.fit)
```

```
##
## Call:
## lm(formula = long_change ~ month + season + nature, data = season_diff)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.30817 -0.03599  0.00530  0.04273  0.50782
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -0.8336700  0.3531651  -2.361   0.0185 *
## monthApril    0.0416468  0.1007984   0.413   0.6796
## monthMay      0.0632772  0.0852459   0.742   0.4582
## monthJune     0.0556884  0.0834504   0.667   0.5048
## monthJuly     0.0361214  0.0831208   0.435   0.6640
## monthAugust   0.0123691  0.0826807   0.150   0.8811
```

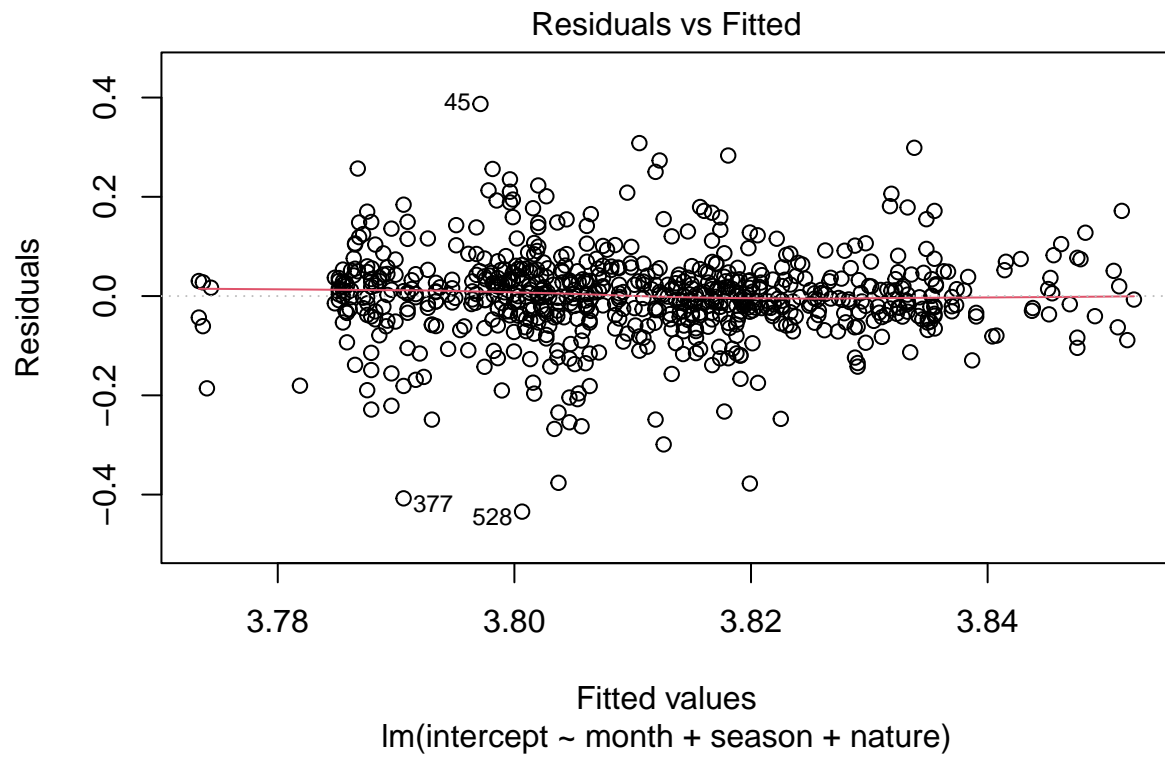
```
## monthSeptember 0.0212965 0.0826071 0.258 0.7966
## monthOctober 0.0341549 0.0826890 0.413 0.6797
## monthNovember 0.0263450 0.0836532 0.315 0.7529
## monthDecember 0.0422468 0.0883312 0.478 0.6326
## season 0.0002184 0.0001715 1.273 0.2033
## natureET -0.0263888 0.0269955 -0.978 0.3287
## natureNR 0.0030556 0.0438742 0.070 0.9445
## natureSS 0.0126339 0.0185659 0.680 0.4964
## natureTS -0.0231521 0.0151062 -1.533 0.1258
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.08164 on 682 degrees of freedom
## Multiple R-squared: 0.05042, Adjusted R-squared: 0.03093
## F-statistic: 2.586 on 14 and 682 DF, p-value: 0.001201
```

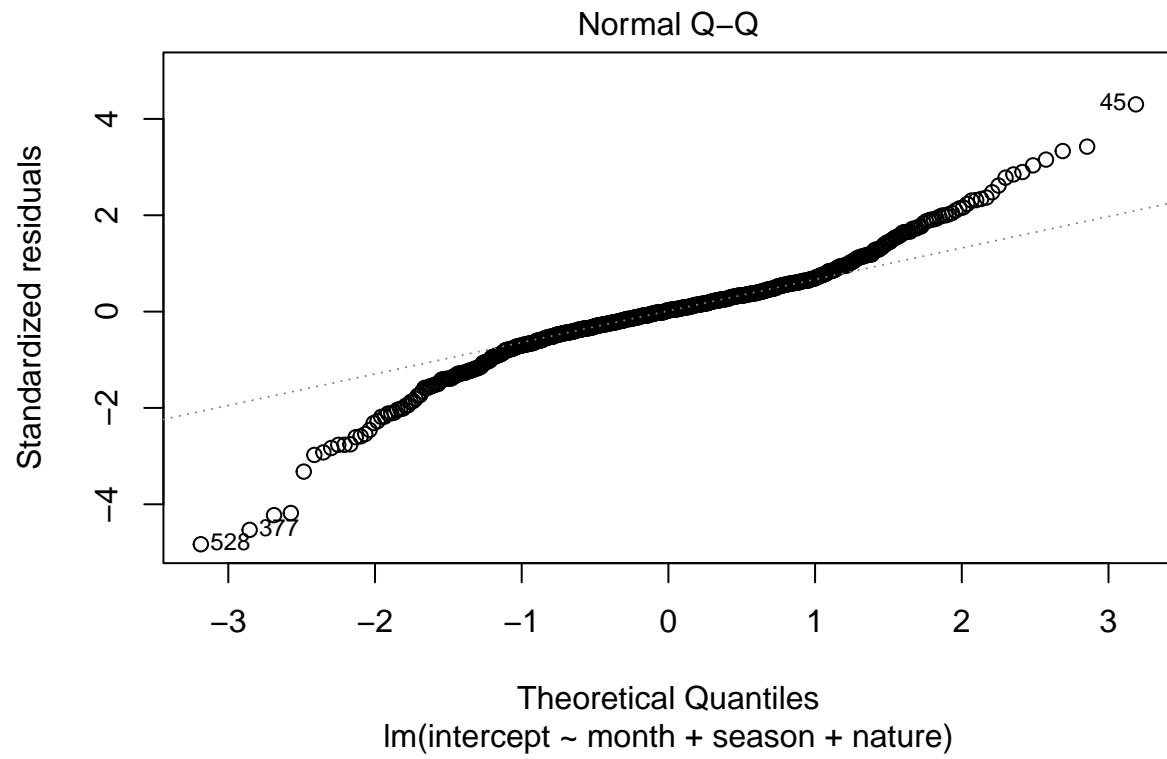
```
summary(wind_change.fit)
```

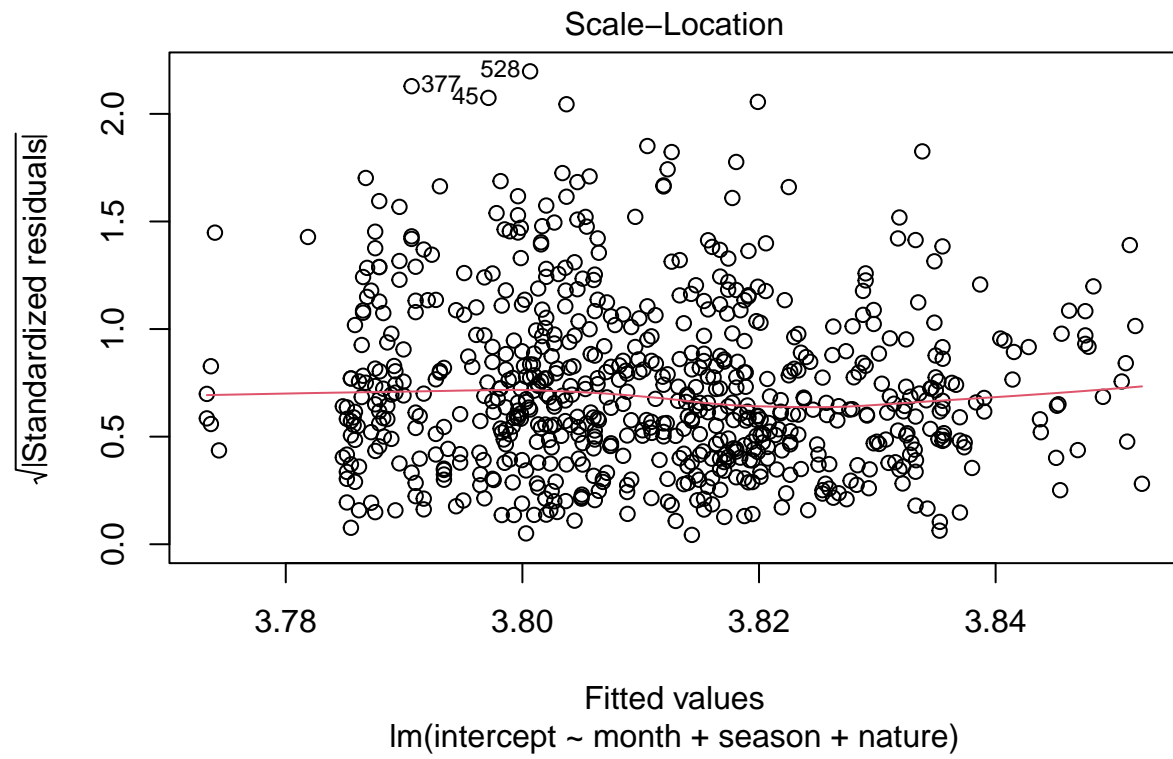
```
##
## Call:
## lm(formula = wind_change ~ month + season + nature, data = season_diff)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.40181 -0.04476 -0.00309  0.04544  0.35691
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  2.890e-01  3.809e-01   0.759   0.448
## monthApril   3.618e-02  1.087e-01   0.333   0.739
## monthMay    -1.629e-02  9.195e-02  -0.177   0.859
## monthJune    2.377e-02  9.001e-02   0.264   0.792
## monthJuly    1.308e-02  8.965e-02   0.146   0.884
## monthAugust  3.124e-02  8.918e-02   0.350   0.726
## monthSeptember 4.448e-02  8.910e-02   0.499   0.618
## monthOctober 3.505e-02  8.919e-02   0.393   0.694
## monthNovember 2.091e-02  9.023e-02   0.232   0.817
## monthDecember 1.142e-02  9.527e-02   0.120   0.905
## season      9.048e-05  1.850e-04   0.489   0.625
## natureET    -2.092e-02  2.912e-02  -0.719   0.473
## natureNR    -2.173e-02  4.732e-02  -0.459   0.646
## natureSS    -2.385e-02  2.003e-02  -1.191   0.234
## natureTS    -1.750e-02  1.629e-02  -1.074   0.283
##
## Residual standard error: 0.08806 on 682 degrees of freedom
## Multiple R-squared: 0.02104, Adjusted R-squared: 0.000943
## F-statistic: 1.047 on 14 and 682 DF, p-value: 0.404
```

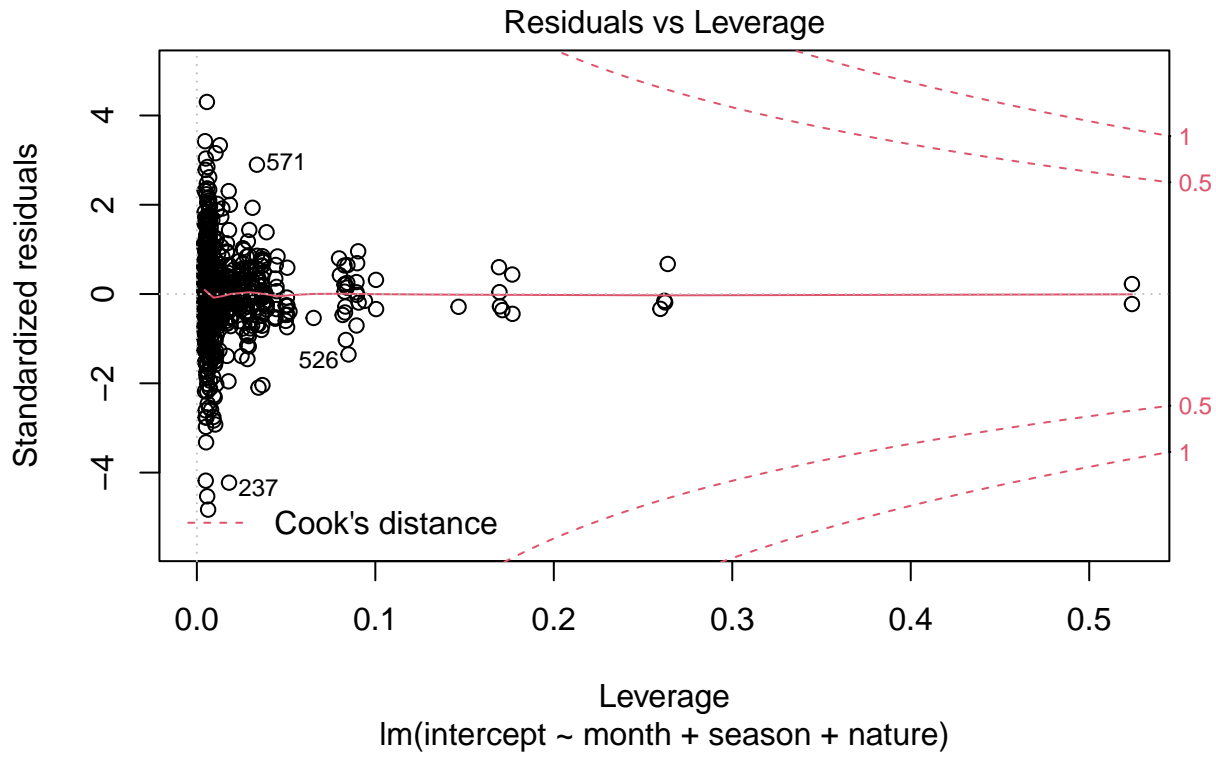
```
plot(intercept.fit)
```

```
## Warning: not plotting observations with leverage one:
##      680
```









```
sum0 <- as.data.frame(summary(intercept.fit)$coefficients[,c(1,4)]) %>% rename(beta0.fit.Est = Estimate)
sum1 <- as.data.frame(summary(wind_prev.fit)$coefficients[,c(1,4)]) %>% rename(beta1.fit.Est = Estimate)
sum2 <- as.data.frame(summary(lat_change.fit)$coefficients[,c(1,4)]) %>% rename(beta2.fit.Est = Estimate)
sum3 <- as.data.frame(summary(long_change.fit)$coefficients[,c(1,4)]) %>% rename(beta3.fit.Est = Estimate)
sum4 <- as.data.frame(summary(wind_change.fit)$coefficients[,c(1,4)]) %>% rename(beta4.fit.Est = Estimate)

kable(cbind(sum0, sum1, sum2, sum3, sum4)) %>%
  kable_paper()
```



	beta0.fit.Est	Pr(> t )	beta1.fit.Est	Pr(> t )	beta2.fit.Est	Pr(> t )	beta3.fit.Est	Pr(> t )
(Intercept)	4.4810021	0.0000000	1.3431063	0.0000000	0.0413063	0.9506172	-0.8336700	0.0118370
monthApril	0.0232609	0.8346449	0.0147943	0.6696787	0.0165579	0.9306863	0.0416468	0.6711111
monthMay	0.0259813	0.7827813	-0.0001180	0.9967888	0.0708822	0.6597505	0.0632772	0.4511111
monthJune	0.0275693	0.7650618	0.0053935	0.8509869	-0.0070875	0.9641298	0.0556884	0.5011111
monthJuly	0.0125400	0.8914489	0.0154032	0.5901741	-0.0090910	0.9538180	0.0361214	0.6611111
monthAugust	-0.0198034	0.8284715	0.0233206	0.4124181	-0.0522548	0.7378961	0.0123691	0.8811111
monthSeptember	-0.0070528	0.9384385	0.0261005	0.3585599	-0.0361073	0.8169707	0.0212965	0.7911111
monthOctober	0.0093435	0.9185853	0.0210829	0.4587183	-0.0286163	0.8546050	0.0341549	0.6711111
monthNovember	0.0145692	0.8748155	0.0246144	0.3925264	0.0239972	0.8792681	0.0263450	0.7511111
monthDecember	0.0057977	0.9526542	0.0088244	0.7715305	-0.0543131	0.7447475	0.0422468	0.6311111
season	-0.0003419	0.0717253	-0.0002252	0.0001471	0.0000365	0.9101708	0.0002184	0.2011111
natureET	0.0008449	0.9774141	0.0037334	0.6877086	-0.0702038	0.1687975	-0.0263888	0.3211111
natureNR	0.0008122	0.9866387	-0.0146142	0.3331114	0.0058967	0.9432660	0.0030556	0.9411111
natureSS	0.0141564	0.4904257	-0.0033299	0.6021721	-0.0013517	0.9692484	0.0126339	0.4911111
natureTS	0.0118370	0.4785102	-0.0059979	0.2486925	-0.0154533	0.5880814	-0.0231521	0.1211111