

Linear model analysis for temperature difference size

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```
library(viridisLite)
```

Temperature difference.

```
d2$dtemp <- d2$Itemp - d2$Otemp
```

```
fit1 <- lm(Itemp ~ poly(glorad, 3) + poly(Otemp, 3) + poly(wv2, 3), data = d2)
fit2 <- lm(dtemp ~ poly(glorad, 3) + poly(Otemp, 3) + poly(wv2, 3), data = d2)
```

Check results

```
summary(fit1)
```

```
##
## Call:
## lm(formula = Itemp ~ poly(glorad, 3) + poly(Otemp, 3) + poly(wv2,
##      3), data = d2)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -7.8227 -0.5742  0.0287  0.5749  7.5529
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    15.07577    0.01404 1073.801 < 2e-16 ***
## poly(glorad, 3)1  47.94051    1.59253   30.103 < 2e-16 ***
## poly(glorad, 3)2  -6.16850    1.24507   -4.954 7.42e-07 ***
## poly(glorad, 3)3   5.05989    1.21319    4.171 3.07e-05 ***
## poly(Otemp, 3)1  472.01562    1.38080  341.841 < 2e-16 ***
## poly(Otemp, 3)2   62.94090    1.23672   50.894 < 2e-16 ***
## poly(Otemp, 3)3   37.86700    1.22338   30.953 < 2e-16 ***
## poly(wv2, 3)1    -0.85992    1.41834   -0.606 0.544345
## poly(wv2, 3)2    -1.13793    1.25875   -0.904 0.366016
## poly(wv2, 3)3    -4.73877    1.22117   -3.881 0.000105 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.195 on 7230 degrees of freedom
## Multiple R-squared:  0.9607, Adjusted R-squared:  0.9606
## F-statistic: 1.963e+04 on 9 and 7230 DF, p-value: < 2.2e-16
```

```
summary(fit2)
```

```
##
## Call:
## lm(formula = dtemp ~ poly(glorad, 3) + poly(0temp, 3) + poly(wv2,
##      3), data = d2)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -7.8227 -0.5742  0.0287  0.5749  7.5529
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      1.11794    0.01404   79.627 < 2e-16 ***
## poly(glorad, 3)1  47.94051    1.59253   30.103 < 2e-16 ***
## poly(glorad, 3)2  -6.16850    1.24507   -4.954 7.42e-07 ***
## poly(glorad, 3)3   5.05989    1.21319    4.171 3.07e-05 ***
## poly(0temp, 3)1   80.78311    1.38080   58.504 < 2e-16 ***
## poly(0temp, 3)2   62.94090    1.23672   50.894 < 2e-16 ***
## poly(0temp, 3)3   37.86700    1.22338   30.953 < 2e-16 ***
## poly(wv2, 3)1     -0.85992    1.41834   -0.606 0.544345
## poly(wv2, 3)2     -1.13793    1.25875   -0.904 0.366016
## poly(wv2, 3)3     -4.73877    1.22117   -3.881 0.000105 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.195 on 7230 degrees of freedom
## Multiple R-squared:  0.6411, Adjusted R-squared:  0.6406
## F-statistic: 1435 on 9 and 7230 DF, p-value: < 2.2e-16
```

Temperature difference model is better—lower residual standard error. (R-squared is lower but that is just because we have already removed a lot of the variation by calculating a difference.)

Let's generate scaled predictor variables for standardized coefficients (relative to 1 standard deviation of predictor variable). This will show which predictors are the most important compared to how much they vary.

```
fit3 <- lm(dtemp ~ poly(scale(glorad), 3) + poly(scale(0temp), 3) + poly(scale(wv2), 3), data = d2)
```

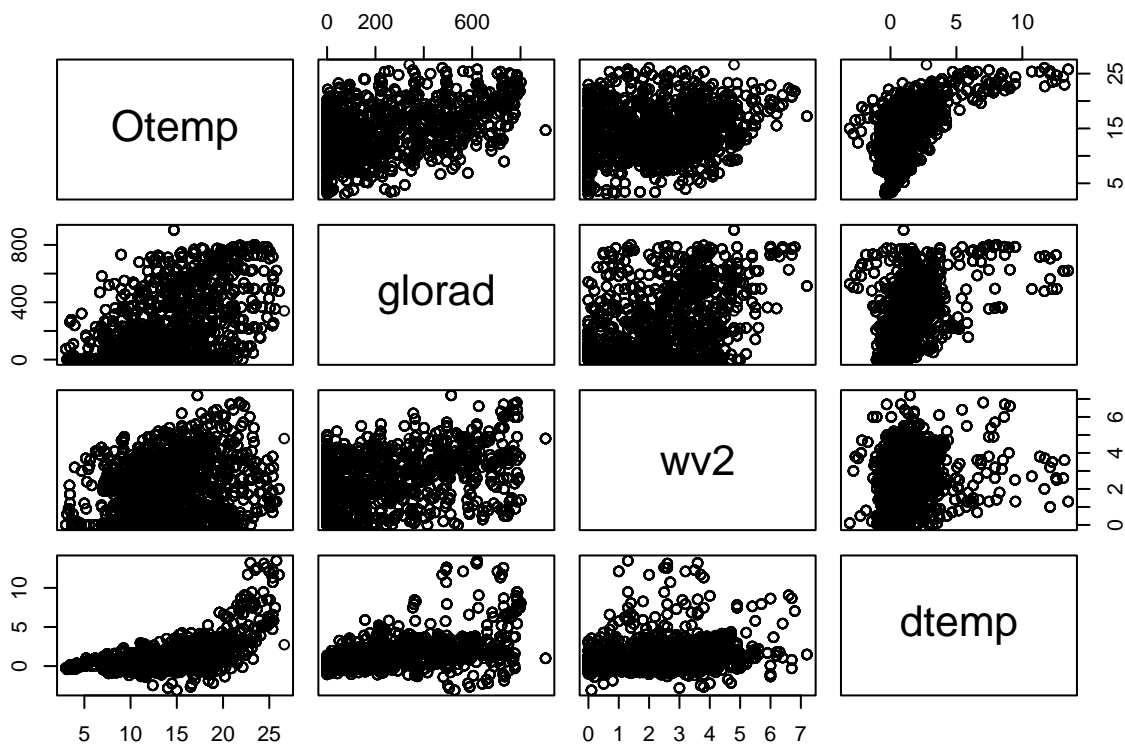
```
summary(fit3)
```

```
##
## Call:
## lm(formula = dtemp ~ poly(scale(glorad), 3) + poly(scale(0temp),
##      3) + poly(scale(wv2), 3), data = d2)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -7.8227 -0.5742  0.0287  0.5749  7.5529
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
```

```
## (Intercept)          1.11794      0.01404  79.627 < 2e-16 ***
## poly(scale(glorad), 3)1 47.94051      1.59253  30.103 < 2e-16 ***
## poly(scale(glorad), 3)2 -6.16850      1.24507  -4.954 7.42e-07 ***
## poly(scale(glorad), 3)3  5.05989      1.21319   4.171 3.07e-05 ***
## poly(scale(Otemp), 3)1  80.78311      1.38080  58.504 < 2e-16 ***
## poly(scale(Otemp), 3)2  62.94090      1.23672  50.894 < 2e-16 ***
## poly(scale(Otemp), 3)3  37.86700      1.22338  30.953 < 2e-16 ***
## poly(scale(wv2), 3)1    -0.85992      1.41834  -0.606 0.544345
## poly(scale(wv2), 3)2    -1.13793      1.25875  -0.904 0.366016
## poly(scale(wv2), 3)3    -4.73877      1.22117  -3.881 0.000105 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.195 on 7230 degrees of freedom
## Multiple R-squared:  0.6411, Adjusted R-squared:  0.6406
## F-statistic: 1435 on 9 and 7230 DF, p-value: < 2.2e-16
```

It looks like temperature (Otemp) is the most important. Is that supported by the measurements?

```
pairs(d2[, .(Otemp, glorad, wv2, dtemp)])
```



Seem so, yes.

Let's see how much worse the model is without the other two.

```
fit4 <- lm(dtemp ~ poly(0temp, 3), data = d2)
```

```
summary(fit4)
```

```
##
## Call:
## lm(formula = dtemp ~ poly(0temp, 3), data = d2)
##
## Residuals:
```

	Min	1Q	Median	3Q	Max
	-8.2102	-0.7387	-0.0205	0.5493	7.8603

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

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	1.11794	0.01519	73.60	<2e-16 ***
poly(0temp, 3)1	103.12430	1.29249	79.79	<2e-16 ***
poly(0temp, 3)2	69.13284	1.29249	53.49	<2e-16 ***
poly(0temp, 3)3	35.25896	1.29249	27.28	<2e-16 ***

```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.292 on 7236 degrees of freedom
## Multiple R-squared:  0.5795, Adjusted R-squared:  0.5793
## F-statistic: 3324 on 3 and 7236 DF, p-value: < 2.2e-16
```

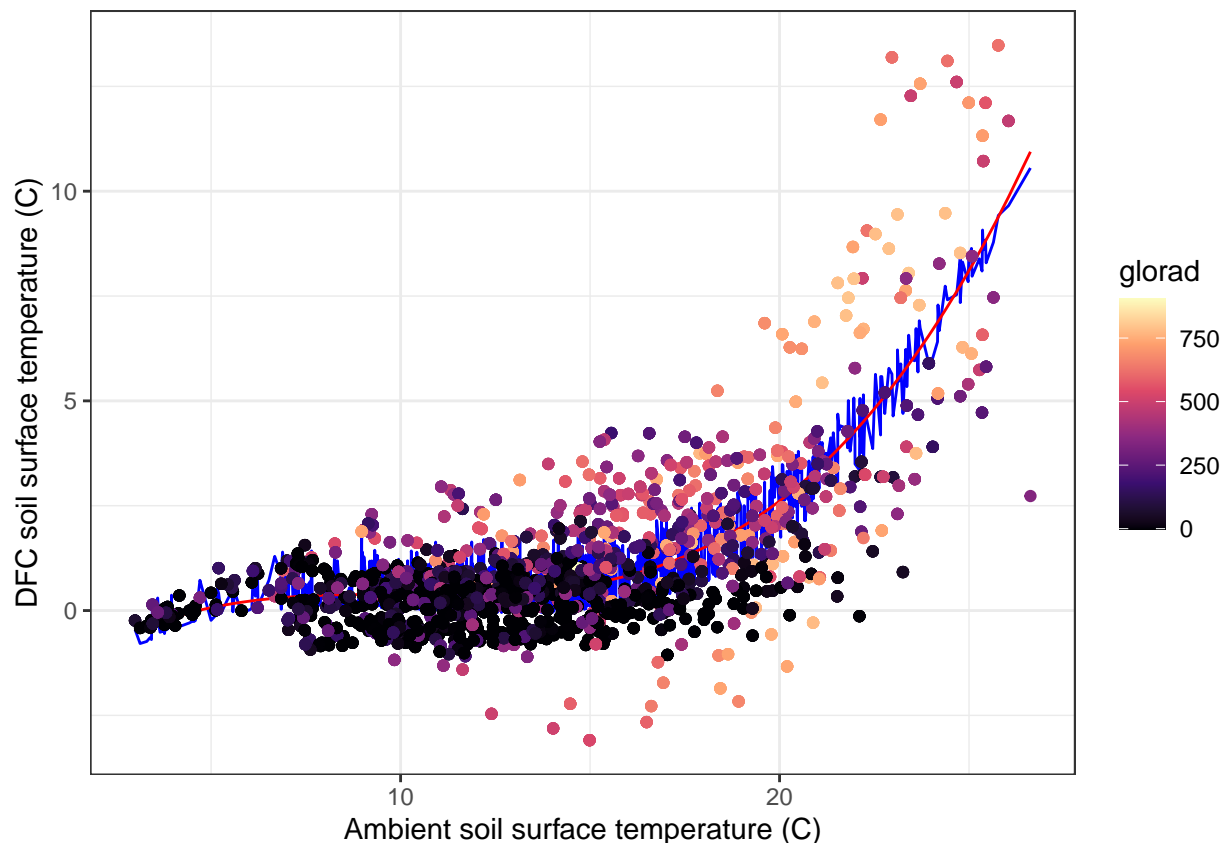
It is almost the same as 2. So effects of radiation and wind look small, surprisingly.

Generate predictions for plotting.

```
d2$Itemp.pred <- predict(fit1)
d2$dtemp.pred2 <- predict(fit2)
d2$dtemp.pred4 <- predict(fit4)
```

And take a look.

```
ggplot(d2, aes(0temp, dtemp, colour = gloriad)) +
  geom_line(aes(0temp, dtemp.pred2, colour = 'blue')) +
  geom_line(aes(0temp, dtemp.pred4, colour = 'red')) +
  geom_point() +
  scale_color_viridis_c(option = 'magma') +
  theme_bw() +
  xlab('Ambient soil surface temperature (C)') + ylab('DFC soil surface temperature (C)')
```



So, temperature alone indeed does as well as the most complete model. But both miss a lot of the variation. How about effects of earlier weather? For that we need to add lagged predictor variables. Try previous hour.

```
wthr <- unique(d2[, .(t.start, Otemp, glorad, wv2)])
wthr[, t.start := t.start - 3600]
d2 <- merge(d2, wthr, by = 't.start', suffixes = c('', '.lag1'))
wthr[, t.start := t.start - 3600]
d2 <- merge(d2, wthr, by = 't.start', suffixes = c('', '.lag2'))
```

```
fit5 <- lm(dtemp ~ poly(glorad, 3) + poly(Otemp, 3) + poly(wv2, 3) + poly(glorad.lag1, 3) + poly(Otemp.lag1, 3) + poly(wv2.lag1, 3) + poly(glorad.lag2, 3) + poly(Otemp.lag2, 3) + poly(wv2.lag2, 3), data = d2)
summary(fit5)
```

```
##
## Call:
## lm(formula = dtemp ~ poly(glorad, 3) + poly(Otemp, 3) + poly(wv2, 3) + poly(glorad.lag1, 3) + poly(Otemp.lag1, 3) + poly(wv2.lag1, 3) + poly(glorad.lag2, 3) + poly(Otemp.lag2, 3) + poly(wv2.lag2, 3), data = d2)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.2062 -0.5393  0.0354  0.5314  6.6285
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      1.13238    0.01311  86.350  < 2e-16 ***
```

```
## poly(glorad, 3)1      95.80272      3.89008  24.627 < 2e-16 ***
## poly(glorad, 3)2     -16.59054      2.00486  -8.275 < 2e-16 ***
## poly(glorad, 3)3       4.82763      1.39403   3.463 0.000537 ***
## poly(Otemp, 3)1     116.77069      6.91358  16.890 < 2e-16 ***
## poly(Otemp, 3)2      -8.41262      4.28916  -1.961 0.049875 *
## poly(Otemp, 3)3     -11.16063      3.15370  -3.539 0.000404 ***
## poly(wv2, 3)1       19.06307      3.15036   6.051 1.51e-09 ***
## poly(wv2, 3)2        4.55615      2.04077   2.233 0.025609 *
## poly(wv2, 3)3       -0.40015      1.57065  -0.255 0.798913
## poly(glorad.lag1, 3)1 -42.18464      3.57016 -11.816 < 2e-16 ***
## poly(glorad.lag1, 3)2  -8.54178      1.91974  -4.449 8.74e-06 ***
## poly(glorad.lag1, 3)3   0.33860      1.37231   0.247 0.805121
## poly(Otemp.lag1, 3)1 -44.78975      7.42097  -6.036 1.66e-09 ***
## poly(Otemp.lag1, 3)2   73.11074      4.36503  16.749 < 2e-16 ***
## poly(Otemp.lag1, 3)3   54.94227      3.20825  17.125 < 2e-16 ***
## poly(wv2.lag1, 3)1   -24.15847      3.18324  -7.589 3.63e-14 ***
## poly(wv2.lag1, 3)2    -4.35724      2.04170  -2.134 0.032866 *
## poly(wv2.lag1, 3)3    -6.75304      1.58355  -4.264 2.03e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.107 on 7111 degrees of freedom
## Multiple R-squared:  0.6951, Adjusted R-squared:  0.6944
## F-statistic: 900.8 on 18 and 7111 DF, p-value: < 2.2e-16
```

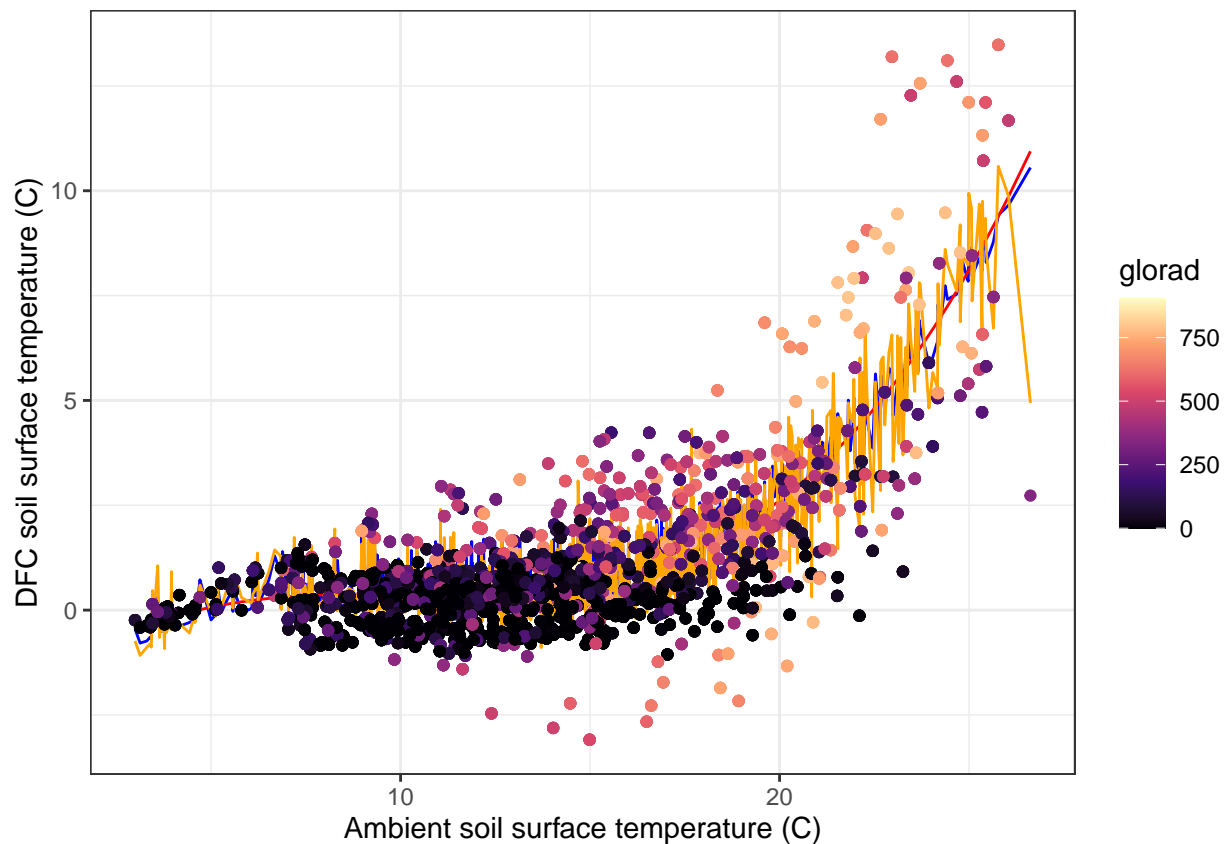
```
summary(fit2)
```

```
##
## Call:
## lm(formula = dtemp ~ poly(glorad, 3) + poly(Otemp, 3) + poly(wv2,
##      3), data = d2)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -7.8227 -0.5742  0.0287  0.5749  7.5529
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    1.11794    0.01404   79.627 < 2e-16 ***
## poly(glorad, 3)1 47.94051    1.59253   30.103 < 2e-16 ***
## poly(glorad, 3)2 -6.16850    1.24507   -4.954 7.42e-07 ***
## poly(glorad, 3)3  5.05989    1.21319    4.171 3.07e-05 ***
## poly(Otemp, 3)1  80.78311    1.38080   58.504 < 2e-16 ***
## poly(Otemp, 3)2  62.94090    1.23672   50.894 < 2e-16 ***
## poly(Otemp, 3)3  37.86700    1.22338   30.953 < 2e-16 ***
## poly(wv2, 3)1   -0.85992    1.41834   -0.606 0.544345
## poly(wv2, 3)2   -1.13793    1.25875   -0.904 0.366016
## poly(wv2, 3)3   -4.73877    1.22117   -3.881 0.000105 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.195 on 7230 degrees of freedom
## Multiple R-squared:  0.6411, Adjusted R-squared:  0.6406
## F-statistic: 1435 on 9 and 7230 DF, p-value: < 2.2e-16
```

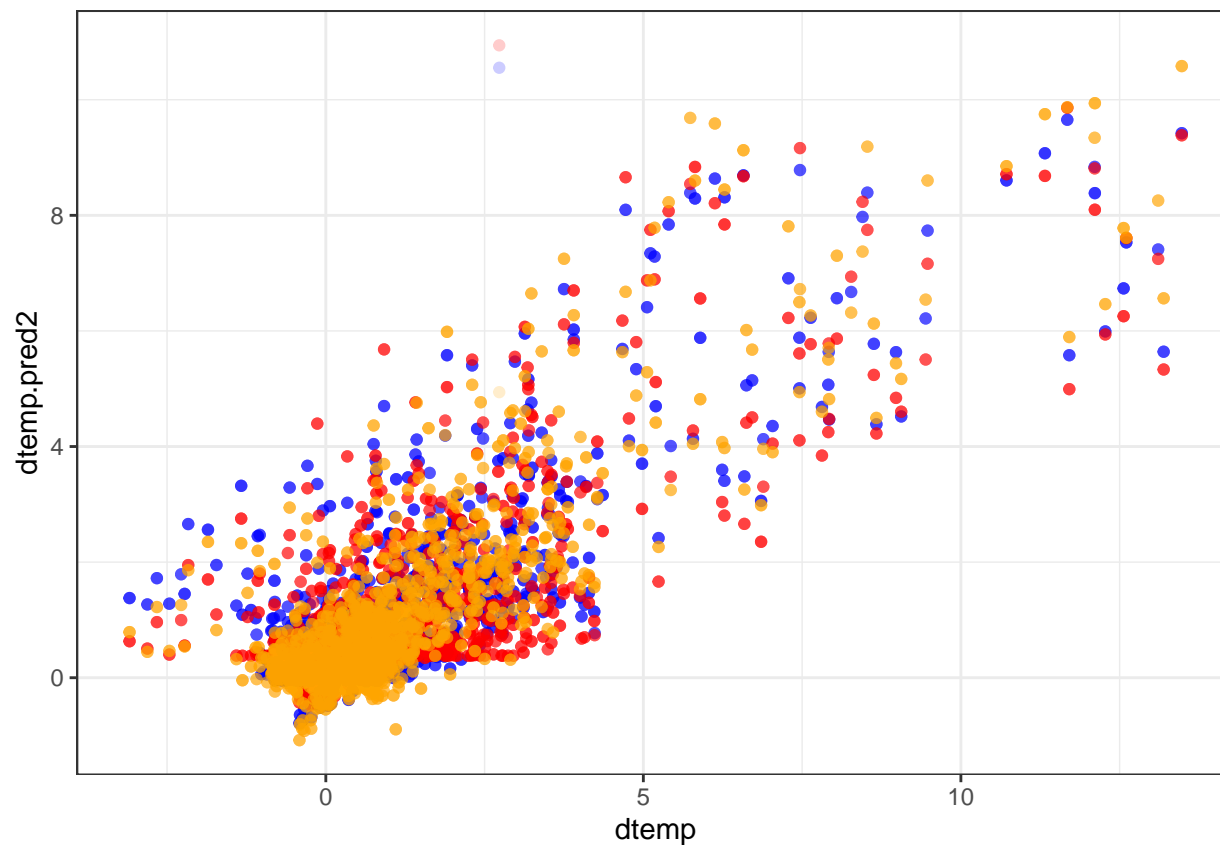
Not a lot of improvement really.

```
d2$dtemp.pred5 <- predict(fit5)
```

```
ggplot(d2, aes(0temp, dtemp, colour = glorad)) +  
  geom_line(aes(0temp, dtemp.pred2, colour = 'blue')) +  
  geom_line(aes(0temp, dtemp.pred4, colour = 'red')) +  
  geom_line(aes(0temp, dtemp.pred5, colour = 'orange')) +  
  geom_point() +  
  scale_color_viridis_c(option = 'magma') +  
  theme_bw() +  
  xlab('Ambient soil surface temperature (C)') + ylab('DFC soil surface temperature (C)')
```



```
ggplot(d2) +  
  geom_point(aes(dtemp, dtemp.pred2, colour = 'blue', alpha = 0.2)) +  
  geom_point(aes(dtemp, dtemp.pred4, colour = 'red', alpha = 0.2)) +  
  geom_point(aes(dtemp, dtemp.pred5, colour = 'orange', alpha = 0.2)) +  
  theme_bw()
```



```
newdat <- expand.grid(glorad = 0:10 * 100, Otemp = 0:10 * 3, wv2 = 2)
setDT(newdat)
newdat$dtemp.pred5 <- predict(fit2, newdata = newdat)
newdat[dtemp.pred5 > 5, ]
```

##	glorad	Otemp	wv2	dtemp.pred5
##	<num>	<num>	<num>	<num>
## 1:	1000	21	2	5.332794
## 2:	0	24	2	5.366631
## 3:	100	24	2	5.784040
## 4:	200	24	2	6.079520
## 5:	300	24	2	6.289129
## 6:	400	24	2	6.448925
## 7:	500	24	2	6.594966
## 8:	600	24	2	6.763309
## 9:	700	24	2	6.990013
## 10:	800	24	2	7.311135
## 11:	900	24	2	7.762733
## 12:	1000	24	2	8.380865
## 13:	0	27	2	10.241125
## 14:	100	27	2	10.658533
## 15:	200	27	2	10.954013
## 16:	300	27	2	11.163623
## 17:	400	27	2	11.323419
## 18:	500	27	2	11.469460


```
## 19:    600    27    2  11.637803
## 20:    700    27    2  11.864507
## 21:    800    27    2  12.185629
## 22:    900    27    2  12.637227
## 23:   1000    27    2  13.255359
## 24:     0    30    2  17.356997
## 25:    100    30    2  17.774406
## 26:    200    30    2  18.069886
## 27:    300    30    2  18.279495
## 28:    400    30    2  18.439291
## 29:    500    30    2  18.585332
## 30:    600    30    2  18.753675
## 31:    700    30    2  18.980379
## 32:    800    30    2  19.301501
## 33:    900    30    2  19.753099
## 34:   1000    30    2  20.371231
##      glorad 0temp    wv2 dtemp.pred5
```

Looks at some subsets.

```
newdat[dtemp.pred5 > 3, ]
```

```
##      glorad 0temp    wv2 dtemp.pred5
##      <num> <num> <num>      <num>
## 1:   1000     9     2   3.105617
## 2:   1000    15     2   3.056096
## 3:    900    18     2   3.078057
## 4:   1000    18     2   3.696189
## 5:    200    21     2   3.031449
## 6:    300    21     2   3.241058
## 7:    400    21     2   3.400854
## 8:    500    21     2   3.546895
## 9:    600    21     2   3.715238
## 10:   700    21     2   3.941942
## 11:   800    21     2   4.263064
## 12:   900    21     2   4.714662
## 13:  1000    21     2   5.332794
## 14:     0    24     2   5.366631
## 15:   100    24     2   5.784040
## 16:   200    24     2   6.079520
## 17:   300    24     2   6.289129
## 18:   400    24     2   6.448925
## 19:   500    24     2   6.594966
## 20:   600    24     2   6.763309
## 21:   700    24     2   6.990013
## 22:   800    24     2   7.311135
## 23:   900    24     2   7.762733
## 24:  1000    24     2   8.380865
## 25:     0    27     2  10.241125
## 26:   100    27     2  10.658533
## 27:   200    27     2  10.954013
## 28:   300    27     2  11.163623
## 29:   400    27     2  11.323419
```

```
## 30:    500    27    2  11.469460
## 31:    600    27    2  11.637803
## 32:    700    27    2  11.864507
## 33:    800    27    2  12.185629
## 34:    900    27    2  12.637227
## 35:   1000    27    2  13.255359
## 36:     0    30    2  17.356997
## 37:    100    30    2  17.774406
## 38:    200    30    2  18.069886
## 39:    300    30    2  18.279495
## 40:    400    30    2  18.439291
## 41:    500    30    2  18.585332
## 42:    600    30    2  18.753675
## 43:    700    30    2  18.980379
## 44:    800    30    2  19.301501
## 45:    900    30    2  19.753099
## 46:   1000    30    2  20.371231
##      glorad 0temp    wv2 dtemp.pred5
```

```
newdat[dtemp.pred5 > 5, ]
```

```
##      glorad 0temp    wv2 dtemp.pred5
##      <num> <num> <num>      <num>
## 1:   1000    21    2    5.332794
## 2:     0    24    2    5.366631
## 3:    100    24    2    5.784040
## 4:    200    24    2    6.079520
## 5:    300    24    2    6.289129
## 6:    400    24    2    6.448925
## 7:    500    24    2    6.594966
## 8:    600    24    2    6.763309
## 9:    700    24    2    6.990013
## 10:   800    24    2    7.311135
## 11:   900    24    2    7.762733
## 12:  1000    24    2    8.380865
## 13:     0    27    2   10.241125
## 14:    100    27    2   10.658533
## 15:    200    27    2   10.954013
## 16:    300    27    2   11.163623
## 17:    400    27    2   11.323419
## 18:    500    27    2   11.469460
## 19:    600    27    2   11.637803
## 20:    700    27    2   11.864507
## 21:    800    27    2   12.185629
## 22:    900    27    2   12.637227
## 23:   1000    27    2   13.255359
## 24:     0    30    2   17.356997
## 25:    100    30    2   17.774406
## 26:    200    30    2   18.069886
## 27:    300    30    2   18.279495
## 28:    400    30    2   18.439291
## 29:    500    30    2   18.585332
## 30:    600    30    2   18.753675
## 31:    700    30    2   18.980379
```

```
## 32:      800      30      2  19.301501
## 33:      900      30      2  19.753099
## 34:     1000      30      2  20.371231
##      glorad 0temp      wv2 dtemp.pred5
```

```
newdat[dtemp.pred5 > 8, ]
```

```
##      glorad 0temp      wv2 dtemp.pred5
##      <num> <num> <num>      <num>
## 1:      1000      24      2    8.380865
## 2:         0      27      2   10.241125
## 3:       100      27      2   10.658533
## 4:       200      27      2   10.954013
## 5:       300      27      2   11.163623
## 6:       400      27      2   11.323419
## 7:       500      27      2   11.469460
## 8:       600      27      2   11.637803
## 9:       700      27      2   11.864507
## 10:      800      27      2   12.185629
## 11:      900      27      2   12.637227
## 12:     1000      27      2   13.255359
## 13:         0      30      2   17.356997
## 14:       100      30      2   17.774406
## 15:       200      30      2   18.069886
## 16:       300      30      2   18.279495
## 17:       400      30      2   18.439291
## 18:       500      30      2   18.585332
## 19:       600      30      2   18.753675
## 20:       700      30      2   18.980379
## 21:       800      30      2   19.301501
## 22:       900      30      2   19.753099
## 23:     1000      30      2   20.371231
##      glorad 0temp      wv2 dtemp.pred5
```