

Prediction of pre-university scores

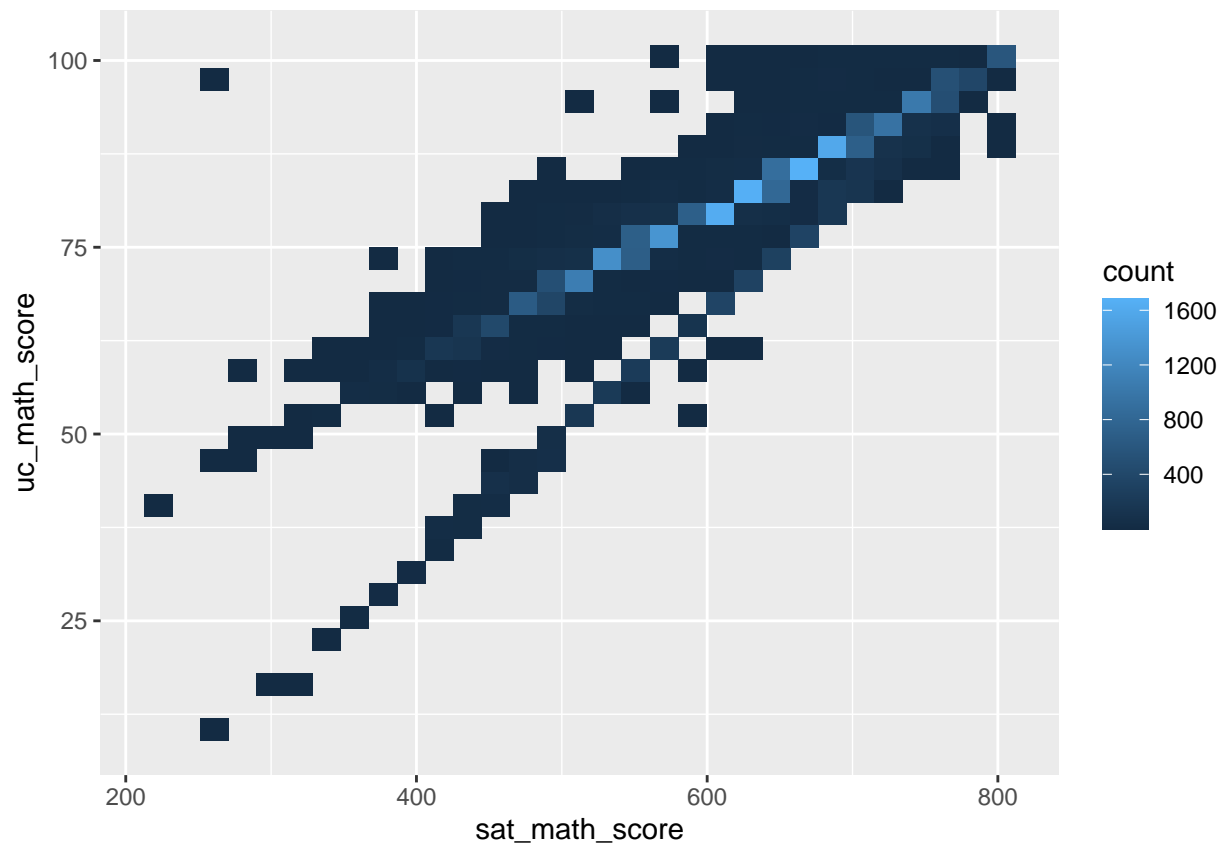
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```
student_background_data = get_student_background_data()
student_sub = get_student_sub()
bg_sub = student_background_data[student_background_data$mellon_id %in% student_sub$mellon_id,]
```

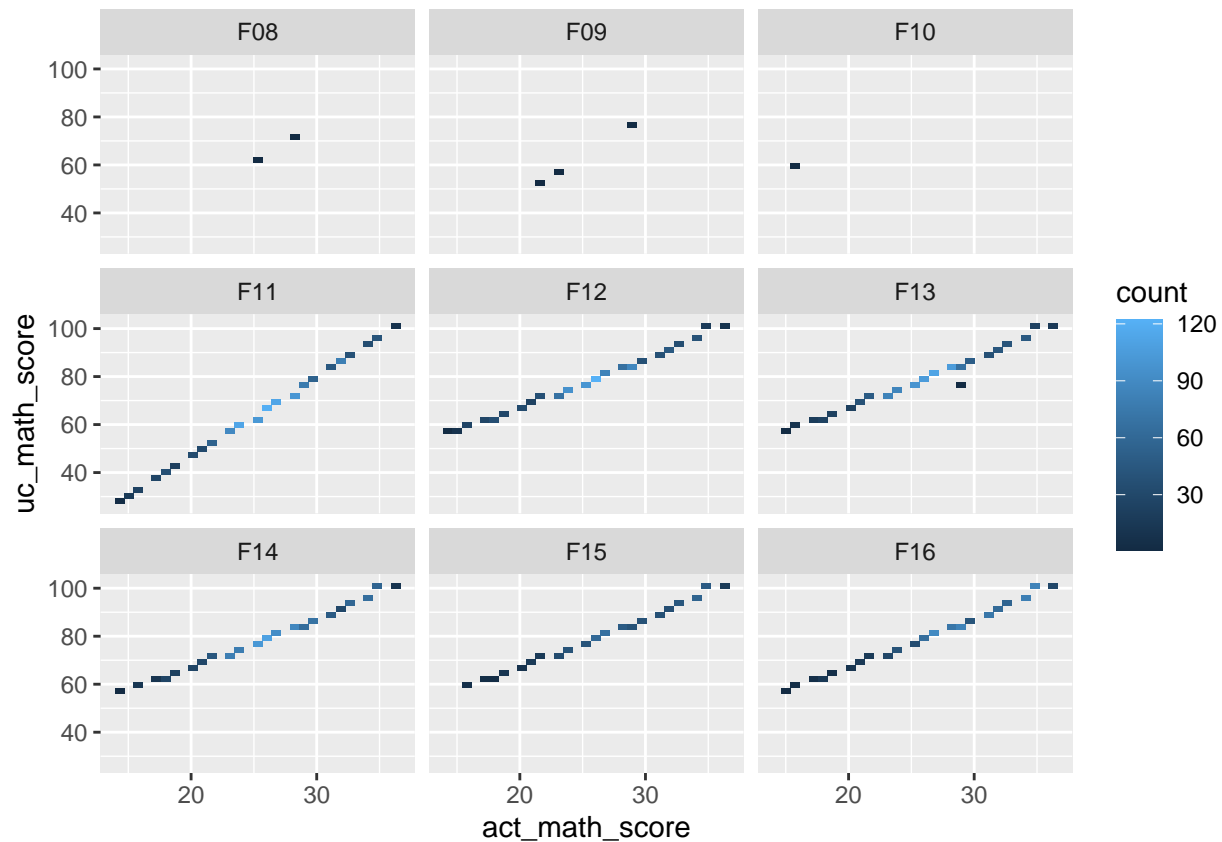
Plot UC math scores for ACT scores

```
ggplot(bg_sub, aes(x=sat_math_score, y=uc_math_score)) +
  geom_bin2d(na.rm=T)
```



Lots of noise and maybe two different mappings. Let's filter out SAT score entries and plot by admission.

```
ggplot(bg_sub[is.na(bg_sub$sat_math_score),], aes(x=act_math_score, y=uc_math_score)) +
  geom_bin2d(na.rm=T) +
  facet_wrap(vars(admitdate))
```



It looks like there was a change in the mapping starting from admitdate Fall 2012.

```
filter_pre_F12 = as.integer(substr(bg_sub$admitdate, 2,3))<12
act_only_F11 = bg_sub[filter_pre_F12&is.na(bg_sub$sat_math_score),]
table(act_only_F11$act_math_score, act_only_F11$uc_math_score)
```

```
##
##      27  30  33  37  40  43  47  50  53  57  60  63  67  70  73  77  80  83
## 14    3   0   0   0   0   0   0   0   0   0   0   0   0   0   0   0   0
## 15    0  17   0   0   0   0   0   0   0   0   0   0   0   0   0   0   0
## 16    0   0  24   0   0   0   0   0   0   0   1   0   0   0   0   0   0
## 17    0   0   0  27   0   0   0   0   0   0   0   0   0   0   0   0   0
## 18    0   0   0   0  37   0   0   0   0   0   0   0   0   0   0   0   0
## 19    0   0   0   0   0  22   0   0   0   0   0   0   0   0   0   0   0
## 20    0   0   0   0   0   0  33   0   0   0   0   0   0   0   0   0   0
## 21    0   0   0   0   0   0   0  31   0   0   0   0   0   0   0   0   0
## 22    0   0   0   0   0   0   0   0  55   0   0   0   0   0   0   0   0
## 23    0   0   0   0   0   0   0   0   0  89   0   0   0   0   0   0   0
## 24    0   0   0   0   0   0   0   0   0   0 112   0   0   0   0   0   0
## 25    0   0   0   0   0   0   0   0   0   0   0 102   0   0   0   0   0
## 26    0   0   0   0   0   0   0   0   0   0   0   0 120   0   0   0   0
## 27    0   0   0   0   0   0   0   0   0   0   0   0   0 112   0   0   0
```

```
## 28 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 101 0 0 0
## 29 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 78 0 0
## 30 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 55 0
## 31 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 45
## 32 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## 33 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## 34 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## 35 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## 36 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
##
##      86  87  90  93  97 100
## 14  0  0  0  0  0  0
## 15  0  0  0  0  0  0
## 16  0  0  0  0  0  0
## 17  0  0  0  0  0  0
## 18  0  0  0  0  0  0
## 19  0  0  0  0  0  0
## 20  0  0  0  0  0  0
## 21  0  0  0  0  0  0
## 22  0  0  0  0  0  0
## 23  0  0  0  0  0  0
## 24  0  0  0  0  0  0
## 25  0  0  0  0  0  0
## 26  0  0  0  0  0  0
## 27  0  0  0  0  0  0
## 28  0  0  0  0  0  0
## 29  0  0  0  0  0  0
## 30  0  0  0  0  0  0
## 31  0  0  0  0  0  0
## 32  0  80  0  0  0  0
## 33  0  0  39  0  0  0
## 34  0  0  0  35  0  0
## 35  0  0  0  0  41  0
## 36  0  0  0  0  0  9
```

There is a clear mapping for admitdate=F11.

```
filter_post_F12 = as.integer(substr(bg_sub$admitdate, 2,3))>=12
act_only_postF12 = bg_sub[filter_post_F12 & is.na(bg_sub$sat_math_score),]
table(act_only_postF12$act_math_score, act_only_postF12$uc_math_score)
```

```
##
##      40  56  58  60  62  63  65  67  69  70  71  73  75  77  79  81  83  85
## 14  0  3  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
## 15  0  0  12  0  0  0  0  0  0  0  0  0  0  0  0  0  0
## 16  0  0  0  59  0  0  0  0  0  0  0  0  0  0  0  0  0
## 17  0  0  0  0  68  0  0  0  0  0  0  0  0  0  0  0  0
## 18  0  0  0  0  0  106  0  0  0  0  0  0  0  0  0  0  0
## 19  0  0  0  0  0  0  93  0  0  0  0  0  0  0  0  0  0
## 20  0  0  0  0  0  0  0  102  0  0  0  0  0  0  0  0  0
## 21  0  0  0  0  0  0  0  0  122  0  0  0  0  0  0  0  0
## 22  0  0  0  0  0  0  0  0  0  161  0  0  0  0  0  0  0
## 23  0  0  0  0  0  0  0  0  0  0  291  0  0  0  0  0  0
```

```

## 24 0 0 0 0 0 0 0 0 0 0 0 0 0 338 0 0 0 0 0
## 25 0 0 0 0 0 0 0 0 0 0 0 0 0 0 374 0 0 0 0
## 26 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 460 0 0 0
## 27 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 442 0 0
## 28 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 379 0
## 29 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 0 0 0 342
## 30 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## 31 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## 32 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## 33 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## 34 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## 35 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## 36 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
##
##      87  89  92  94  97 100
## 14  0  0  0  0  0  0
## 15  0  0  0  0  0  0
## 16  0  0  0  0  0  0
## 17  0  0  0  0  0  0
## 18  0  0  0  0  0  0
## 19  0  0  0  0  0  0
## 20  0  0  0  0  0  0
## 21  0  0  0  0  0  0
## 22  0  0  0  0  0  0
## 23  0  0  0  0  0  0
## 24  0  0  0  0  0  0
## 25  0  0  0  0  0  0
## 26  0  0  0  0  0  0
## 27  0  0  0  0  0  0
## 28  0  0  0  0  0  0
## 29  0  0  0  0  0  0
## 30 238  0  0  0  0
## 31  0 235  0  0  0
## 32  0  0 215  0  0
## 33  0  0  0 234  0
## 34  0  0  0  0 285  0
## 35  0  0  0  0  0 228
## 36  0  0  0  0  0  82

```

Let's create the mapping.

```

uc_scores_11 = unique(act_only_F11$uc_math_score)
uc_scores_11 = uc_scores_11[!is.na(uc_scores_11)][-24] # remove 86
uc_scores_post12 = unique(act_only_postF12$uc_math_score)
uc_scores_post12 = uc_scores_post12[!is.na(uc_scores_post12)]
uc_scores_post12 = c(uc_scores_post12[order(uc_scores_post12)][-c(1, 10)], 100) # remove 40, 70, add 100
uc_act = data.frame(act=14:36, uc_pre_12=uc_scores_11[order(uc_scores_11)], uc_post_12=uc_scores_post12)
uc_act

```

```

##   act uc_pre_12 uc_post_12
## 1   14         27         56
## 2   15         30         58
## 3   16         33         60

```

```
## 4 17 37 62
## 5 18 40 63
## 6 19 43 65
## 7 20 47 67
## 8 21 50 69
## 9 22 53 71
## 10 23 57 73
## 11 24 60 75
## 12 25 63 77
## 13 26 67 79
## 14 27 70 81
## 15 28 73 83
## 16 29 77 85
## 17 30 80 87
## 18 31 83 89
## 19 32 87 92
## 20 33 90 94
## 21 34 93 97
## 22 35 97 100
## 23 36 100 100
```

```
lm(uc_pre_12~act,uc_act)
```

```
##
## Call:
## lm(formula = uc_pre_12 ~ act, data = uc_act)
##
## Coefficients:
## (Intercept)          act
##      -19.953         3.332
```

```
uc_act$uc_pre_12 == round(uc_act$act*10/3-20)
```

```
## [1] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
## [16] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
```

```
lm(uc_post_12~act,uc_act)
```

```
##
## Call:
## lm(formula = uc_post_12 ~ act, data = uc_act)
##
## Coefficients:
## (Intercept)          act
##      26.632         2.036
```

```
uc_act$uc_post_12 == round(uc_act$act*2.036+26.63)
```

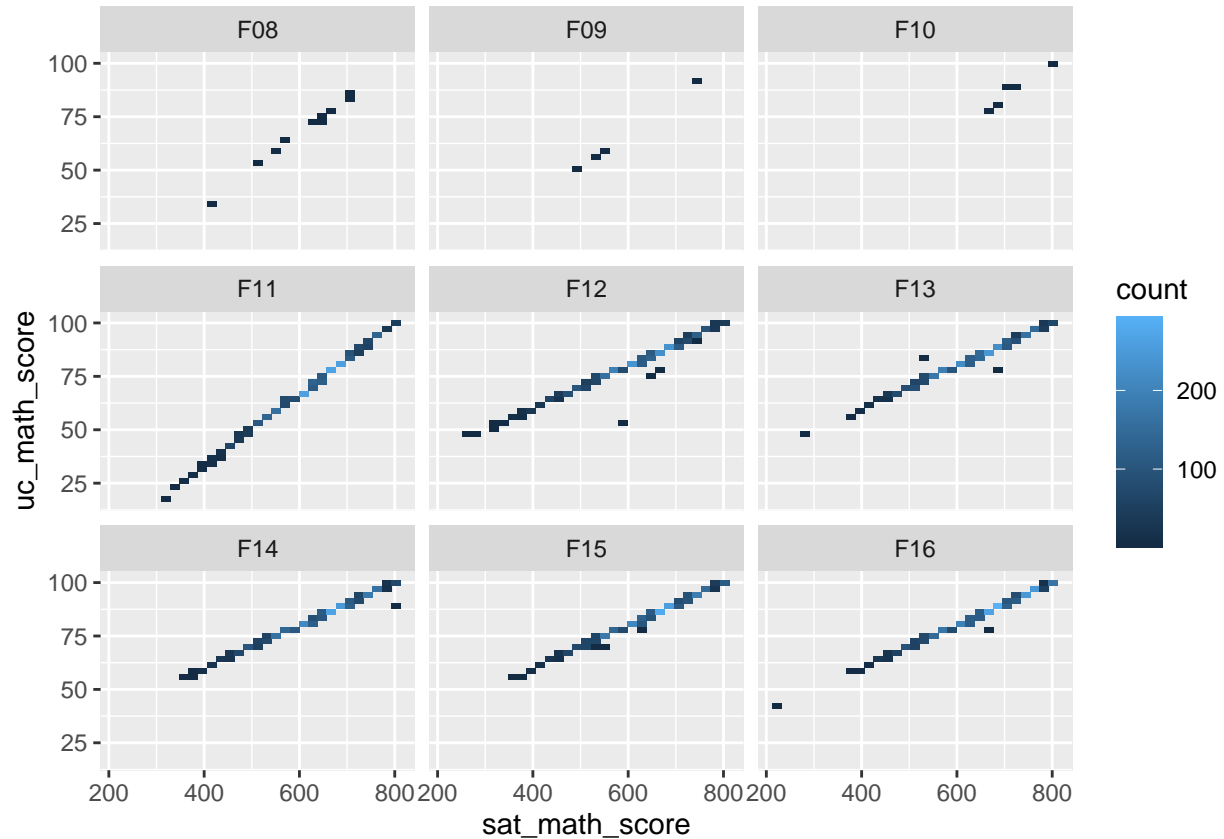
```
## [1] FALSE FALSE FALSE FALSE TRUE TRUE TRUE TRUE TRUE TRUE TRUE FALSE
## [13] FALSE FALSE FALSE FALSE FALSE FALSE TRUE TRUE FALSE FALSE TRUE
```

This does not look linear enough. Let's then print the mapping for the config.

```
## [1] "data.frame(act=c(14,15,16,17,18,19,20,21,22,23,24,25,26,27,28,29,30,31,32,33,34,35,36),uc_pre_12=
```

Plot SAT to UC math score by admission

```
ggplot(bg_sub[is.na(bg_sub$act_math_score),], aes(x=sat_math_score, y=uc_math_score)) +
  geom_bin2d(na.rm=T) +
  facet_wrap(vars(admitdate))
```



Just a little bit of noise but also a change between 11 and 12. So, let's also separate for the admit data to find the mapping.

```
options(max.print=3000)
sat_only_pre_F12 = bg_sub[filter_pre_F12&is.na(bg_sub$act_math_score),]
mappings = table(sat_only_pre_F12$sat_math_score, sat_only_pre_F12$uc_math_score)
mappings
```

```
##
##      18  23  25  27  28  30  32  33  35  37  38  40  42  43  45  47  48  50
## 300    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 310    2    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 330    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 340    0    2    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 350    0    0    4    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 360    0    0    0    5    0    0    0    0    0    0    0    0    0    0    0    0    0
## 370    0    0    0    0    5    0    0    0    0    0    0    0    0    0    0    0    0
## 380    0    0    0    0    0    4    0    0    0    0    0    0    0    0    0    0    0
## 390    0    0    0    0    0    0    5    0    0    0    0    0    0    0    0    0    0
```

##	400	0	0	0	0	0	0	0	6	0	0	0	0	0	0	0	0	0
##	410	0	0	0	0	0	0	0	0	12	0	0	0	0	0	0	0	0
##	420	0	0	0	0	0	0	0	0	0	15	0	0	0	0	0	0	0
##	430	0	0	0	0	0	0	0	0	0	0	11	0	0	0	0	0	0
##	440	0	0	0	0	0	0	0	0	0	0	0	16	0	0	0	0	0
##	450	0	0	0	0	0	0	0	0	0	0	0	0	19	0	0	0	0
##	460	0	0	0	0	0	0	0	0	0	0	0	0	0	42	0	0	0
##	470	0	0	0	0	0	0	0	0	0	0	0	0	0	0	34	0	0
##	480	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	26	0
##	490	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	39
##	500	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	39
##	510	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	520	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	530	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	540	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	550	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	560	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	570	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	580	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	590	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	600	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	610	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	620	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	630	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	640	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	650	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	660	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	670	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	680	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	690	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	700	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	710	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	720	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	730	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	740	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	750	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	760	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	770	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	780	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	790	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	800	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##																		
##		52	53	55	57	58	60	62	63	65	67	68	70	72	73	75	77	78
##	300	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	310	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	330	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	340	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	350	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	360	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	370	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	380	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	390	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	400	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	410	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0

##	420	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	430	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	440	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	450	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	460	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	470	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	480	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	490	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	500	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	510	50	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	520	0	58	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	530	0	0	48	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	540	0	0	0	88	0	0	0	0	0	0	0	0	0	0	0	0	0
##	550	0	0	0	0	64	0	0	0	0	0	0	0	0	0	0	0	0
##	560	0	0	0	0	0	97	0	0	0	0	0	0	0	0	0	0	0
##	570	0	0	0	0	0	0	76	0	0	0	0	0	0	0	0	0	0
##	580	0	0	0	0	0	0	0	79	0	0	0	0	0	0	0	0	0
##	590	0	0	0	0	0	0	0	0	80	0	0	0	0	0	0	0	0
##	600	0	0	0	0	0	0	0	0	0	103	0	0	0	0	0	0	0
##	610	0	0	0	0	0	0	0	0	0	0	157	0	0	0	0	0	0
##	620	0	0	0	0	0	0	0	0	0	0	0	112	0	0	0	0	0
##	630	0	0	0	0	0	0	0	0	0	0	0	0	131	0	0	0	0
##	640	0	0	0	0	0	0	0	0	0	0	0	0	0	123	0	0	0
##	650	0	0	0	0	0	0	0	0	0	0	0	0	0	0	113	0	0
##	660	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	137	0
##	670	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	130
##	680	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	127
##	690	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	700	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	710	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	720	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	730	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	740	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	750	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	760	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	770	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	780	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	790	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	800	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##																		
##		82	83	85	86	87	88	90	92	93	95	97	98	100				
##	300	0	0	0	0	0	0	0	0	0	0	0	0	0				
##	310	0	0	0	0	0	0	0	0	0	0	0	0	0				
##	330	0	0	0	0	0	0	0	0	0	0	0	0	0				
##	340	0	0	0	0	0	0	0	0	0	0	0	0	0				
##	350	0	0	0	0	0	0	0	0	0	0	0	0	0				
##	360	0	0	0	0	0	0	0	0	0	0	0	0	0				
##	370	0	0	0	0	0	0	0	0	0	0	0	0	0				
##	380	0	0	0	0	0	0	0	0	0	0	0	0	0				
##	390	0	0	0	0	0	0	0	0	0	0	0	0	0				
##	400	0	0	0	0	0	0	0	0	0	0	0	0	0				
##	410	0	0	0	0	0	0	0	0	0	0	0	0	0				
##	420	0	0	0	0	0	0	0	0	0	0	0	0	0				
##	430	0	0	0	0	0	0	0	0	0	0	0	0	0				


```

## 440 0 0 0 0 0 0 0 0 0 0 0 0 0
## 450 0 0 0 0 0 0 0 0 0 0 0 0 0
## 460 0 0 0 0 0 0 0 0 0 0 0 0 0
## 470 0 0 0 0 0 0 0 0 0 0 0 0 0
## 480 0 0 0 0 0 0 0 0 0 0 0 0 0
## 490 0 0 0 0 0 0 0 0 0 0 0 0 0
## 500 0 0 0 0 0 0 0 0 0 0 0 0 0
## 510 0 0 0 0 0 0 0 0 0 0 0 0 0
## 520 0 0 0 0 0 0 0 0 0 0 0 0 0
## 530 0 0 0 0 0 0 0 0 0 0 0 0 0
## 540 0 0 0 0 0 0 0 0 0 0 0 0 0
## 550 0 0 0 0 0 0 0 0 0 0 0 0 0
## 560 0 0 0 0 0 0 0 0 0 0 0 0 0
## 570 0 0 0 0 0 0 0 0 0 0 0 0 0
## 580 0 0 0 0 0 0 0 0 0 0 0 0 0
## 590 0 0 0 0 0 0 0 0 0 0 0 0 0
## 600 0 0 0 0 0 0 0 0 0 0 0 0 0
## 610 0 0 0 0 0 0 0 0 0 0 0 0 0
## 620 0 0 0 0 0 0 0 0 0 0 0 0 0
## 630 0 0 0 0 0 0 0 0 0 0 0 0 0
## 640 0 0 0 0 0 0 0 0 0 0 0 0 0
## 650 0 0 0 0 0 0 0 0 0 0 0 0 0
## 660 0 0 0 0 0 0 0 0 0 0 0 0 0
## 670 0 0 0 0 0 0 0 0 0 0 0 0 0
## 680 0 0 0 0 0 0 0 0 0 0 0 0 0
## 690 131 0 0 0 0 0 0 0 0 0 0 0 0
## 700 0 112 0 0 0 0 1 0 0 0 0 0 0
## 710 0 0 97 0 0 0 0 0 0 0 0 0 0
## 720 0 0 0 0 45 0 0 0 0 0 0 0 0
## 730 0 0 0 0 0 71 0 0 0 0 0 0 0
## 740 0 0 0 0 0 0 56 0 0 0 0 0 0
## 750 0 0 0 0 0 0 0 68 0 0 0 0 0
## 760 0 0 0 0 0 0 0 0 47 0 0 0 0
## 770 0 0 0 0 0 0 0 0 0 87 0 0 0
## 780 0 0 0 0 0 0 0 0 0 0 25 0 0
## 790 0 0 0 0 0 0 0 0 0 0 0 4 0
## 800 0 0 0 0 0 0 0 0 0 0 0 0 46

```

```

mappings['700','90']=0 # fix the only outlier
uc_sat_pre_12 = data.frame(sat = as.integer(rownames(mappings)[(row(mappings)[which(mappings != 0)])]),
                           uc = as.integer(colnames(mappings)[(col(mappings)[which(mappings != 0)])]))
uc_sat_pre_12

```

```

##   sat  uc
## 1 310 18
## 2 340 23
## 3 350 25
## 4 360 27
## 5 370 28
## 6 380 30
## 7 390 32
## 8 400 33
## 9 410 35
## 10 420 37

```

```
## 11 430 38
## 12 440 40
## 13 450 42
## 14 460 43
## 15 470 45
## 16 480 47
## 17 490 48
## 18 500 50
## 19 510 52
## 20 520 53
## 21 530 55
## 22 540 57
## 23 550 58
## 24 560 60
## 25 570 62
## 26 580 63
## 27 590 65
## 28 600 67
## 29 610 68
## 30 620 70
## 31 630 72
## 32 640 73
## 33 650 75
## 34 660 77
## 35 670 78
## 36 680 80
## 37 690 82
## 38 700 83
## 39 710 85
## 40 720 87
## 41 730 88
## 42 740 90
## 43 750 92
## 44 760 93
## 45 770 95
## 46 780 97
## 47 790 98
## 48 800 100
```

Try to find a formula.

```
lm(uc ~ sat, uc_sat_pre_12)
```

```
##
## Call:
## lm(formula = uc ~ sat, data = uc_sat_pre_12)
##
## Coefficients:
## (Intercept)          sat
##    -33.4138         0.1668
```

```
uc_sat_pre_12$uc == round((uc_sat_pre_12$sat - 200) / 6)
```

```
## [1] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
## [16] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
## [31] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
## [46] TRUE TRUE TRUE
```

```
sat_only_post_F12 = bg_sub[filter_post_F12&is.na(bg_sub$act_math_score),]
mappings = table(sat_only_post_F12$sat_math_score, sat_only_post_F12$uc_math_score)
mappings
```

```
##
##      40  42  47  48  49  51  52  53  54  55  56  57  58  59  60  61  62  63
## 220    0    1    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 270    0    0    1    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 280    0    0    0    3    0    0    0    0    0    0    0    0    0    0    0    0    0
## 290    0    0    0    0    2    0    0    0    0    0    0    0    0    0    0    0    0
## 310    0    0    0    0    0    3    0    0    0    0    0    0    0    0    0    0    0
## 320    0    0    0    0    0    0    1    0    0    0    0    0    0    0    0    0    0
## 330    0    0    0    0    0    0    0    2    0    0    0    0    0    0    0    0    0
## 340    0    0    0    0    0    0    0    0    4    0    0    0    0    0    0    0    0
## 350    0    0    0    0    0    0    0    0    0    5    0    0    0    0    0    0    0
## 360    0    0    0    0    0    0    0    0    0    0    10    0    0    0    0    0    0
## 370    0    0    0    0    0    0    0    0    0    0    0    10    0    0    0    0    0
## 380    0    0    0    0    0    0    0    0    0    0    0    0    17    0    0    0    0
## 390    0    0    0    0    0    0    0    0    0    0    0    0    0    18    0    0    0
## 400    0    0    0    0    0    0    0    0    0    0    0    0    0    0    43    0    0
## 410    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    40    0
## 420    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    51
## 430    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    72
## 440    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 450    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 460    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 470    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 480    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 490    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 500    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 510    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 520    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 530    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 540    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 550    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 560    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 570    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 580    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 590    0    0    0    0    0    0    0    1    0    0    0    0    0    0    0    0    0
## 600    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 610    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 620    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 630    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 640    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
## 650    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0    0
```

##	660	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	670	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	680	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	690	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	700	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	710	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	720	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	730	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	740	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	750	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	760	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	770	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	780	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	790	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	800	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##																			
##		64	65	66	67	68	69	70	71	72	73	74	75	76	77	78	79	80	81
##	220	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	270	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	280	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	290	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	310	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	320	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	330	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	340	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	350	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	360	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	370	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	380	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	390	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	400	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	410	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	420	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	430	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	440	92	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	450	0	99	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	460	0	0	144	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	470	0	0	0	188	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	480	0	0	0	0	171	0	0	0	0	0	0	0	0	0	0	0	0	0
##	490	0	0	0	0	0	199	0	0	0	0	0	0	0	0	0	0	0	0
##	500	0	0	0	0	0	0	258	0	0	0	0	0	0	0	0	0	0	0
##	510	0	0	0	0	0	0	0	324	0	0	0	0	0	0	0	0	0	0
##	520	0	0	0	0	0	0	0	0	300	0	0	0	0	0	0	0	0	0
##	530	0	0	0	0	0	0	0	0	0	355	0	0	0	0	0	0	0	0
##	540	0	0	0	0	0	0	0	1	0	0	348	0	0	0	0	0	0	0
##	550	0	0	0	0	0	1	0	0	0	0	0	393	0	0	0	0	0	0
##	560	0	0	0	0	0	0	0	0	0	0	0	0	426	0	0	0	0	0
##	570	0	0	0	0	0	0	0	0	0	0	0	0	0	405	0	0	0	0
##	580	0	0	0	0	0	0	0	0	0	0	0	0	0	0	430	0	0	0
##	590	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	441	0	0
##	600	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	511	0
##	610	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	554
##	620	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	630	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0

##	640	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	650	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	
##	660	0	0	0	0	0	0	0	0	0	0	0	0	1	0	1	0	0	
##	670	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	680	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	690	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0	
##	700	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	710	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	720	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	730	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	740	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	750	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	760	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	770	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	780	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	790	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##	800	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
##																			
##		82	83	84	85	86	87	88	89	90	91	92	93	94	95	96	97	98	99
##	220	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	270	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	280	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	290	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	310	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	320	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	330	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	340	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	350	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	360	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	370	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	380	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	390	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	400	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	410	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	420	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	430	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	440	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	450	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	460	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	470	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	480	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	490	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	500	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	510	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	520	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	530	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	540	0	0	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	550	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	560	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	570	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	580	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	590	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	600	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	610	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0

##	620	577	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	630	0	566	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	640	0	0	588	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	650	0	0	0	599	0	0	0	0	0	0	0	0	0	0	0	0	0
##	660	0	0	0	0	634	0	0	0	0	0	0	0	0	0	0	0	0
##	670	0	0	0	0	0	639	0	0	0	0	0	0	0	0	0	0	0
##	680	0	0	0	0	0	0	655	0	0	0	0	0	0	0	0	0	0
##	690	0	0	0	0	0	0	0	590	0	0	0	0	0	0	0	0	0
##	700	0	0	0	0	0	0	0	0	582	0	0	0	0	0	0	0	0
##	710	0	0	0	0	0	0	0	0	0	476	0	0	0	0	0	0	0
##	720	0	0	0	0	0	0	0	0	0	0	452	0	0	0	0	0	0
##	730	0	0	0	0	0	0	0	0	0	0	0	324	0	0	0	0	0
##	740	0	0	0	0	0	0	0	0	0	0	0	0	526	0	0	0	0
##	750	0	0	0	0	0	0	0	0	0	0	1	0	0	369	0	0	0
##	760	0	0	0	0	0	0	0	0	0	0	0	0	0	0	349	0	0
##	770	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	446	0
##	780	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	169
##	790	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	143
##	800	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0
##																		
##		100																
##	220	0																
##	270	0																
##	280	0																
##	290	0																
##	310	0																
##	320	0																
##	330	0																
##	340	0																
##	350	0																
##	360	0																
##	370	0																
##	380	0																
##	390	0																
##	400	0																
##	410	0																
##	420	0																
##	430	0																
##	440	0																
##	450	0																
##	460	0																
##	470	0																
##	480	0																
##	490	0																
##	500	0																
##	510	0																
##	520	0																
##	530	0																
##	540	0																
##	550	0																
##	560	0																
##	570	0																
##	580	0																
##	590	0																

```
## 600 0
## 610 0
## 620 0
## 630 0
## 640 0
## 650 0
## 660 0
## 670 0
## 680 0
## 690 0
## 700 0
## 710 0
## 720 0
## 730 0
## 740 0
## 750 0
## 760 0
## 770 0
## 780 0
## 790 0
## 800 459
```

```
# fix outliers
mappings['590','53']=0
mappings['550','69']=0
mappings['540','71']=0
mappings['540','84']=0
mappings['630','78']=0
mappings['650','75']=0
mappings['660','77']=0
mappings['660','79']=0
mappings['690','79']=0
mappings['750','92']=0
mappings['800','88']=0
#mappings
```

```
uc_sat_post_12 = data.frame(sat = as.integer(rownames(mappings)[(row(mappings)[which(mappings != 0)])])
                           uc = as.integer(colnames(mappings)[(col(mappings)[which(mappings != 0)])])
uc_sat_post_12$uc == (uc_sat_post_12$sat + 200) / 10
```

```
## [1] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
## [16] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
## [31] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
## [46] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
```

Prediction of math score from SAT, ACT and Combination

- SAT scores to UC are two affine transformations:

```
print(get_uc_from_sat)
```

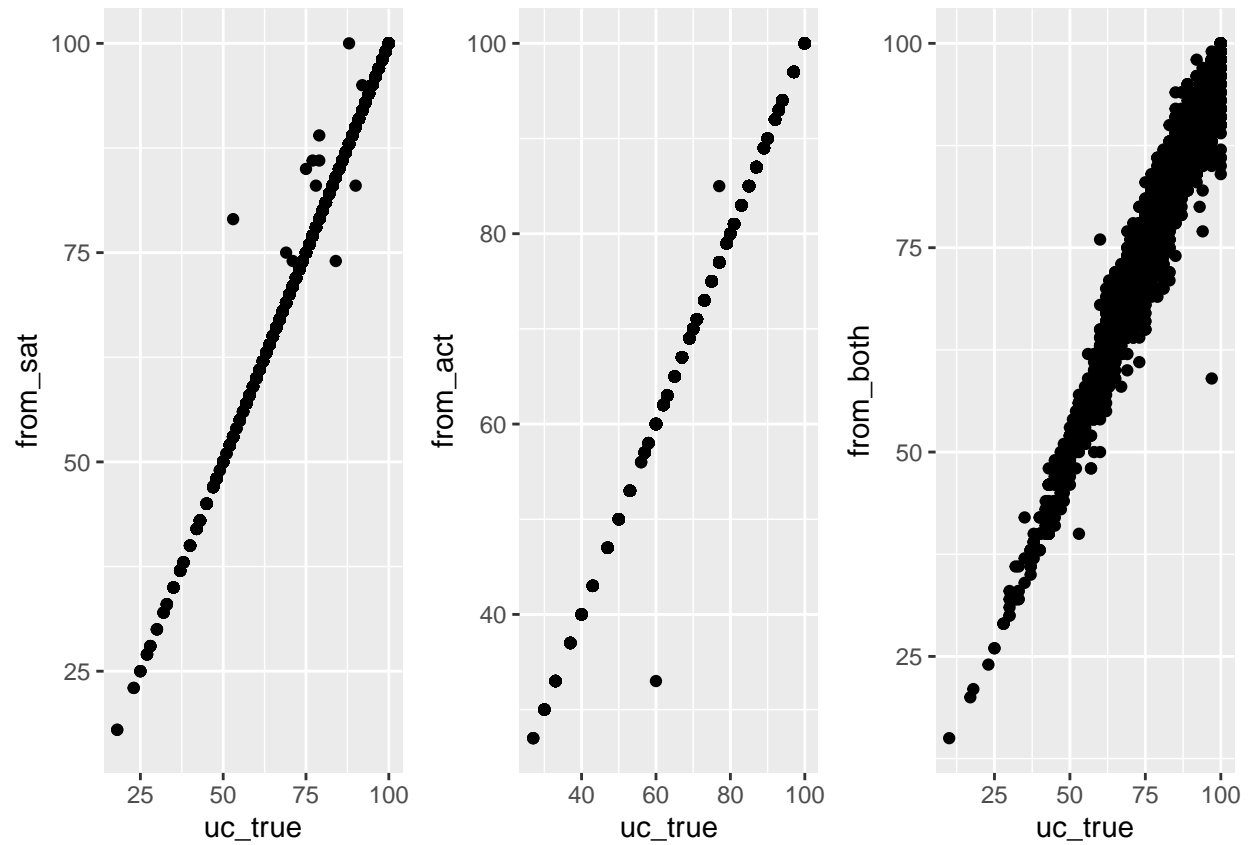
```
## function (sat, pre_12 = F)
## {
##   if (pre_12)
##     return(round((sat - 200)/6))
##   return(round((sat + 200)/10))
## }
```

- ACT follows an almost-linear mapping

```
uc_math_from_sat = mapply(get_uc_from_sat, bg_sub$sat_math_score, filter_pre_F12)
uc_math_from_act = mapply(get_uc_from_act, bg_sub$act_math_score, filter_pre_F12)
uc_math_from_sat_and_act = round(0.75 * uc_math_from_sat + 0.25 * uc_math_from_act)
```

Plot predictions

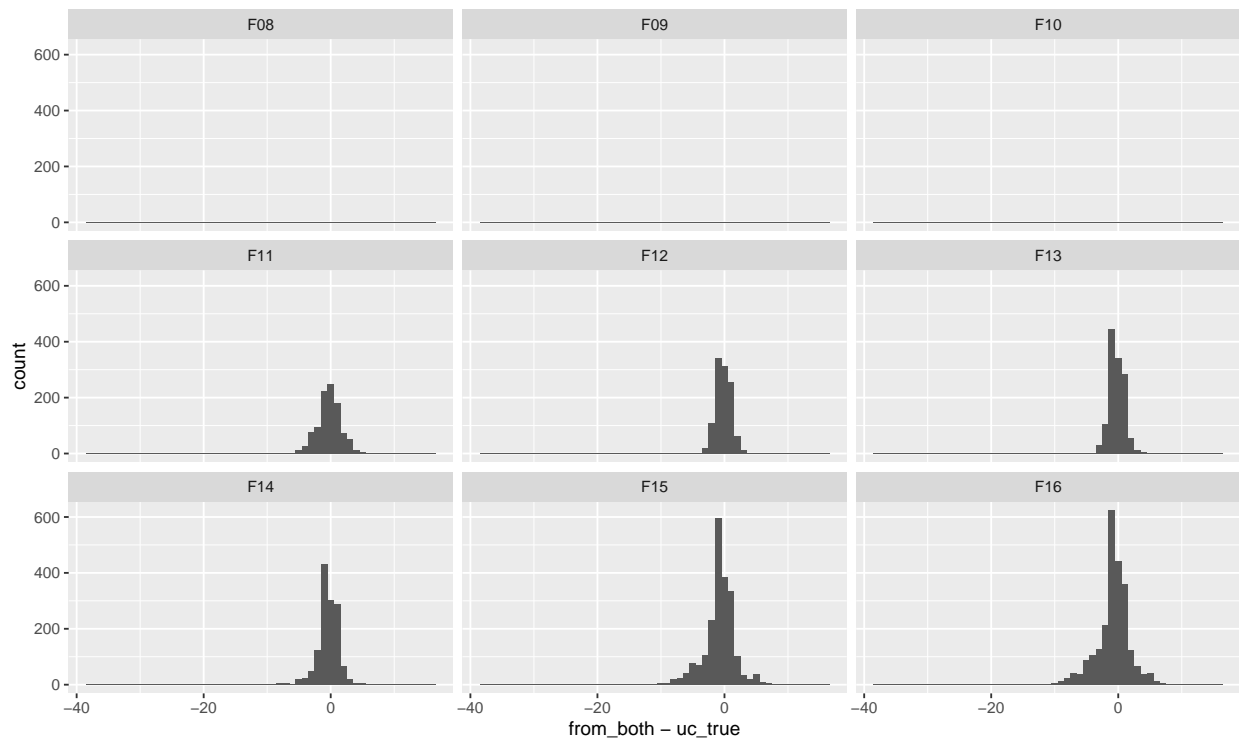
```
uc_math_scores = data.frame(uc_true=bg_sub$uc_math_score,
                             from_sat=uc_math_from_sat,
                             from_act=uc_math_from_act,
                             from_both=uc_math_from_sat_and_act,
                             admitdate=bg_sub$admitdate)
p1 = ggplot(uc_math_scores[is.na(bg_sub$act_math_score),], aes(x=uc_true, y=from_sat)) +
  geom_point(na.rm=T)
p2 = ggplot(uc_math_scores[is.na(bg_sub$sat_math_score),], aes(x=uc_true, y=from_act)) +
  geom_point(na.rm=T)
p3 = ggplot(uc_math_scores, aes(x=uc_true, y=from_both)) +
  geom_point(na.rm=T)
grid.arrange(p1, p2, p3, ncol=3)
```

```
print('Break grid arrange')
```

```
## [1] "Break grid arrange"
```

Plot prediction errors by admission



- combination has some problems, that do not depend on admission year.

```
model = lm(uc_true ~ from_act + from_sat, uc_math_scores)
summary(model)
```

```
##
## Call:
## lm(formula = uc_true ~ from_act + from_sat, data = uc_math_scores)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -16.544  -1.099  -0.101   0.901  37.735
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  0.142748   0.167153   0.854   0.393
## from_act     0.252396   0.004392  57.473 <2e-16 ***
## from_sat     0.753046   0.004557 165.249 <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.166 on 9273 degrees of freedom
## (37132 Beobachtungen als fehlend gelöscht)
## Multiple R-squared:  0.9593, Adjusted R-squared:  0.9592
## F-statistic: 1.092e+05 on 2 and 9273 DF, p-value: < 2.2e-16
```

Let's check predictions

```
table(uc_math_scores$uc - round(predict(model, uc_math_scores)))
```

```
##
##  -17  -10   -9   -8   -7   -6   -5   -4   -3   -2   -1    0    1    2    3    4
##    1    1    4    5   37   42   58  185  389  941 2211 2418 1520  706  248  159
##    5    6    7    8    9   10   11   12   13   14   16   17   38
##  172   53   43   46   15    6    6    2    4    1    1    1    1
```

Although majority is predicted very well, it looks like the formula does not quite get it right. Maybe let's see what information we have in the important three years.

```
table(!is.na(bg_sub$act_math_score[filter_pre_F12]))
```

```
##
## FALSE  TRUE
## 15878  2295
```

```
table(!is.na(bg_sub$sat_math_score[filter_pre_F12]))
```

```
##
## FALSE  TRUE
##  4220 13953
```

```
table(is.na(bg_sub$sat_math_score[filter_pre_F12]) & !is.na(bg_sub$act_math_score[filter_pre_F12]))
```

```
##
## FALSE  TRUE
## 16905  1268
```

We mostly have SAT scores and only 1268 cases where we exclusively know the ACT math score, which means that we might ignore the remaining residuals.

Next steps: - use the prediction for math scores - investigate on meaning of UC score: <https://blog.prepscholar.com/historical-act-percentiles-2015-2014-2013-2012-2011> - check predictiveness of UC scores for dropout