pokemon dataset

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```
library(ggplot2)
  library(dplyr)
Attaching package: 'dplyr'
The following objects are masked from 'package:stats':
    filter, lag
The following objects are masked from 'package:base':
    intersect, setdiff, setequal, union
  library(mltools)
  library(data.table)
Attaching package: 'data.table'
The following objects are masked from 'package:dplyr':
    between, first, last
  library(car)
Loading required package: carData
```

```
Attaching package: 'car'

The following object is masked from 'package:dplyr':
recode
```

The professor recommended doing this so the empty string "" will be converted to a NA value so we can then use the is.na() function.

```
pokemon <- read.csv("/Users/sarah/Desktop/SDS291/FinalProject/pokemon.csv", na.strings = c</pre>
```

Making the dual_type column that says if the Pokemon is a dual type based on if there is a second type.

```
pokemon <- mutate(pokemon, dual_type = !is.na(type2))</pre>
```

Making sure that there are only TRUE and FALSE values in the dual_type column.

```
unique(pokemon$'dual_type')
```

[1] TRUE FALSE

Making sure that these variables are the right data type.

```
pokemon$'capture_rate' = as.numeric(pokemon$'capture_rate')
```

Warning: NAs introduced by coercion

```
pokemon$'type1' = as.factor(pokemon$'type1')
pokemon$'type2' = as.factor(pokemon$'type2')
```

Dataset

We are planning to use the Complete Pokemon Dataset that has information on different Pokemon up to Gen 7. The link where we got the dataset is included below. Dataset Link

Research Question

How do different Pokemon's base stats influence capture rate?

```
pokemon$'capture_rate' = as.numeric(pokemon$'capture_rate')
pokemon$'type1' = as.factor(pokemon$'type1')
pokemon$'type2' = as.factor(pokemon$'type2')
```

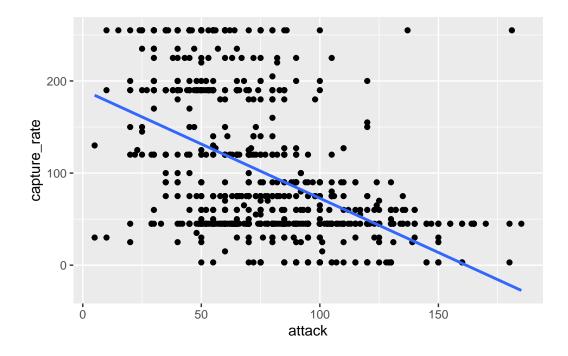
Different Possible Stats

- attack
- base_happiness
- base_egg_steps
- base total
- defense
- hp
- sp_attack
- sp_defense
- speed

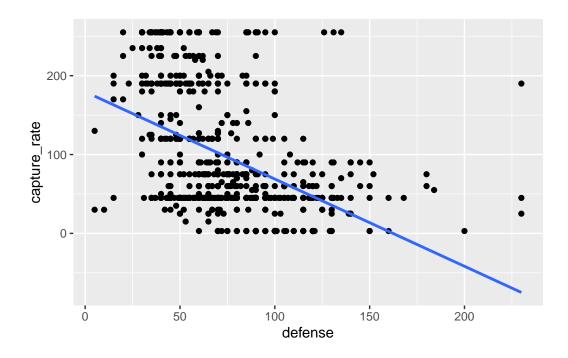
The base stats I'll focus on are attack, hp, defense, and speed. I'm not sure what are important base stats for Pokemon but I'm guessing.

```
ggplot(data = pokemon, mapping = aes(x = attack, y = capture_rate)) +
   geom_point() +
   geom_smooth(method = lm, se = FALSE, formula = y~x)
```

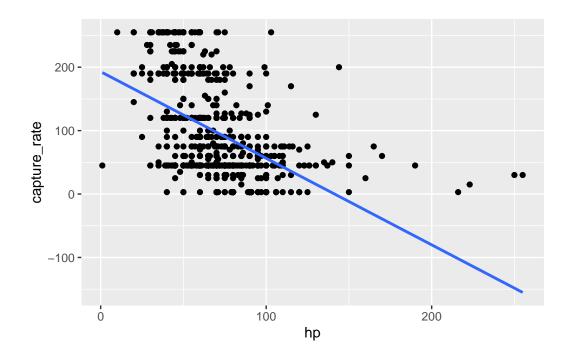
Warning: Removed 1 rows containing non-finite values (stat_smooth).



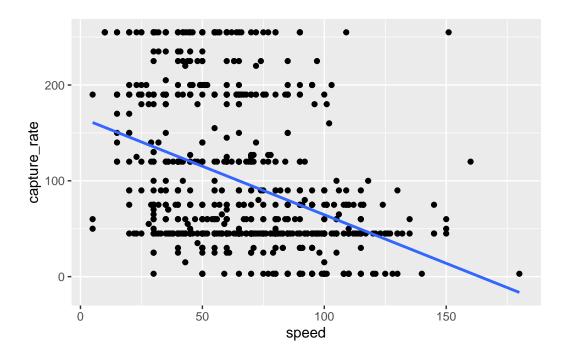
```
ggplot(data = pokemon, mapping = aes(x = defense, y = capture_rate)) +
    geom_point() +
    geom_smooth(method = lm, se = FALSE, formula = y~x)
```



```
ggplot(data = pokemon, mapping = aes(x = hp, y = capture_rate)) +
   geom_point() +
   geom_smooth(method = lm, se = FALSE, formula = y~x)
```



```
ggplot(data = pokemon, mapping = aes(x = speed, y = capture_rate)) +
   geom_point() +
   geom_smooth(method = lm, se = FALSE, formula = y~x)
```



Making the models

Additive Model

```
additive_capture_model2 <- lm(capture_rate ~ attack + defense + hp + speed, data = pokemon summary(additive_capture_model2)
```

Call:

lm(formula = capture_rate ~ attack + defense + hp + speed, data = pokemon)

Residuals:

Min 1Q Median 3Q Max -150.832 -36.856 -5.036 36.345 255.371

Coefficients:

Estimate Std. Error t value Pr(>|t|)
(Intercept) 288.48251 7.67680 37.578 < 2e-16 ***
attack -0.29466 0.08042 -3.664 0.000265 ***

Adjusted R-squared: 0.453

Interactive Model

```
interact_capture_model2 <- lm(capture_rate ~ attack * defense * hp * speed, data = pokemon
summary(interact_capture_model2)</pre>
```

Call:

lm(formula = capture_rate ~ attack * defense * hp * speed, data = pokemon)

Residuals:

Min 1Q Median 3Q Max -167.956 -31.949 -1.533 28.992 219.782

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	2.615e+02	5.582e+01	4.686	3.29e-06 ***
attack	-7.279e-02	9.924e-01	-0.073	0.94154
defense	1.677e-01	5.320e-01	0.315	0.75270
hp	-7.348e-02	8.772e-01	-0.084	0.93326
speed	1.101e+00	9.999e-01	1.101	0.27102
attack:defense	-1.027e-02	1.017e-02	-1.010	0.31278
attack:hp	-1.235e-02	1.323e-02	-0.933	0.35101
defense:hp	-1.291e-02	1.015e-02	-1.273	0.20355
attack:speed	-2.156e-02	1.591e-02	-1.355	0.17572
defense:speed	-3.126e-02	1.115e-02	-2.803	0.00519 **
hp:speed	-2.085e-02	1.636e-02	-1.274	0.20303

```
1.270 0.20436
                        1.798e-04 1.415e-04
attack:defense:hp
                        3.228e-04 1.606e-04
attack:defense:speed
                                               2.010 0.04482 *
attack:hp:speed
                        2.942e-04 2.235e-04
                                               1.316 0.18844
defense:hp:speed
                        2.655e-04 1.861e-04
                                               1.426 0.15422
attack:defense:hp:speed -3.217e-06 2.283e-06 -1.409 0.15925
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 55.09 on 784 degrees of freedom
  (1 observation deleted due to missingness)
Multiple R-squared: 0.488, Adjusted R-squared: 0.4782
F-statistic: 49.81 on 15 and 784 DF, p-value: < 2.2e-16
```

Adjusted R-squared: 0.4782

Nested F-test

The p-value is below 0.05 implying that the change from additive to interactive is necessary, but I feel like the adjusted r-square doesn't justify the change.

Opinion: Would be interesting to look at but haven't tested removing some explanatory variables to see how it affects the model.

Model we plan to use

We plan to use an interaction model to predict the capture rate of pokemon (outcome variable). We plan to use the base stats of the Pokemon as our explanatory variables. We are currently

looking into attack, hp, defense, and speed, but we are thinking of adding base_total as a possible explanatory variable and also seeing if we can add our own variable that says if the Pokemon is a dual type, but are having trouble with creating the variable at the moment.

Current Plan

- Look at how using a different number and combination of variables affect the model
- what stats we plan to look at: dual_type, attack, defense, hp, speed, base_total, base_happiness, sp_attack, and sp_defense. Plan to look at a different combination of these things, but should also look at the VIF of these variables as well to make sure there isn't too much multicollinearity.

VIF and Multicollinearity

All values are below 10 so multicollinearity wont be an issue here.

```
vif(interact_capture_model2)
```

there are higher-order terms (interactions) in this model consider setting type = 'predictor'; see ?vif

attack	defense	hp
268.34144	70.61604	143.24012
speed	attack:defense	attack:hp
219.30342	520.29476	644.42856
defense:hp	attack:speed	defense:speed
275.89995	996.83905	279.76076
hp:speed	attack:defense:hp	attack:defense:speed
572.21339	950.03798	1054.74108
attack:hp:speed	defense:hp:speed	$\verb attack:defense:hp:speed $
1766.61410	753.94466	1918.23200

So all these values are above 10 so they should be centered if we chose to use them.

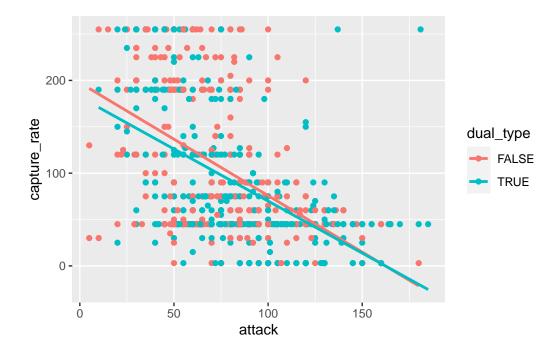
Adding dual types to the current models

So I'll be using attack, speed, hp, defense,and dual_type as explanatory variables to predict capture rate. I'll make the mode, and tey to add to the visualizations but haven't tried using new base stat explanatory variables since I'm not sure if using adjusted r-squared or a nested f-test would be the best way to compare so once we know that we can drop, add,or try new variables a bit better.

Visualizations

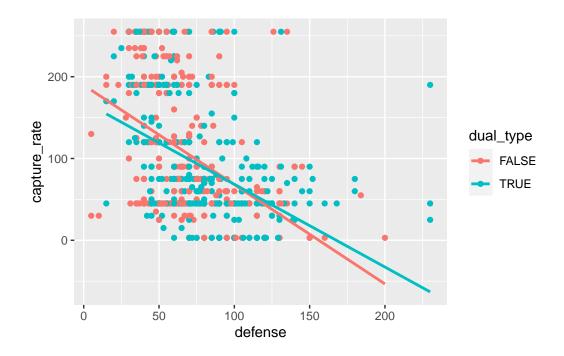
```
ggplot(data = pokemon, mapping = aes(x = attack, y = capture_rate, color=dual_type)) +
    geom_point() +
    geom_smooth(method = lm, se = FALSE, formula = y~x)
```

Warning: Removed 1 rows containing non-finite values (stat_smooth).



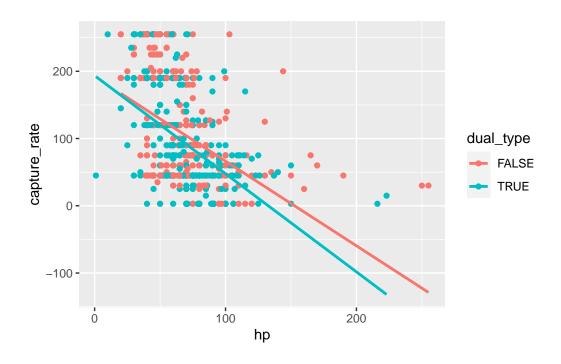
```
ggplot(data = pokemon, mapping = aes(x = defense, y = capture_rate, color=dual_type)) +
   geom_point() +
   geom_smooth(method = lm, se = FALSE, formula = y~x)
```

Warning: Removed 1 rows containing missing values (geom_point).

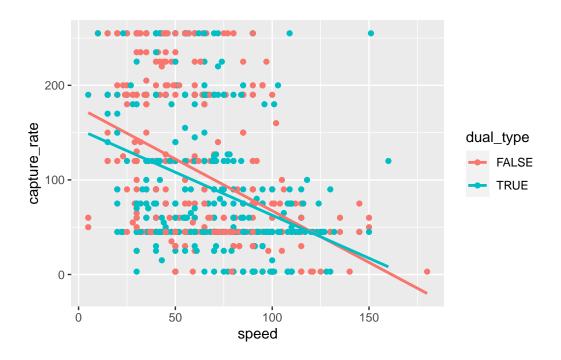


```
ggplot(data = pokemon, mapping = aes(x = hp, y = capture_rate, color=dual_type)) +
    geom_point() +
    geom_smooth(method = lm, se = FALSE, formula = y~x)
```

Warning: Removed 1 rows containing non-finite values (stat_smooth).



```
ggplot(data = pokemon, mapping = aes(x = speed, y = capture_rate, color=dual_type)) +
   geom_point() +
   geom_smooth(method = lm, se = FALSE, formula = y~x)
```



Making the models

Additive Model

```
add_capture_model3 <- lm(capture_rate ~ attack + defense + hp + speed + dual_type, data =
summary(add_capture_model3)</pre>
```

Call:

Residuals:

```
Min 1Q Median 3Q Max -150.647 -36.921 -4.964 36.208 255.453
```

Coefficients:

```
Estimate Std. Error t value Pr(>|t|) (Intercept) 288.55739 7.74097 37.277 < 2e-16 ***
```

```
-0.29463
                         0.08047 -3.661 0.000267 ***
attack
              -0.77304
                         0.07567 -10.217 < 2e-16 ***
defense
              -0.86777
                         0.08256 -10.511 < 2e-16 ***
hp
              -0.76053
                         0.07554 -10.068 < 2e-16 ***
speed
                         4.06062 -0.078 0.937646
dual_typeTRUE -0.31776
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 56.43 on 794 degrees of freedom
  (1 observation deleted due to missingness)
Multiple R-squared: 0.4558, Adjusted R-squared: 0.4523
F-statistic:
             133 on 5 and 794 DF, p-value: < 2.2e-16
```

Adjusted R-squared: 0.4523

Interactive Model

```
interact_capture_model3 <- lm(capture_rate ~ attack * defense * hp * speed * dual_type, da
summary(interact_capture_model3)</pre>
```

Call:

Residuals:

```
Min 1Q Median 3Q Max
-180.69 -31.66 -2.61 30.32 216.10
```

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	1.605e+02	9.847e+01	1.630	0.1035
attack	2.804e+00	1.601e+00	1.752	0.0802 .
defense	9.493e-01	1.494e+00	0.635	0.5254
hp	-7.295e-02	1.408e+00	-0.052	0.9587
speed	1.957e+00	1.692e+00	1.157	0.2477
dual_typeTRUE	1.818e+02	1.382e+02	1.316	0.1886
attack:defense	-3.548e-02	2.065e-02	-1.718	0.0862 .
attack:hp	-2.844e-02	1.927e-02	-1.476	0.1404

```
-1.052e-02 2.184e-02 -0.482
defense:hp
                                                                  0.6303
                                   -6.283e-02 2.559e-02 -2.455
attack:speed
                                                                  0.0143 *
defense:speed
                                   -2.841e-02 2.636e-02 -1.078
                                                                  0.2815
                                   -1.533e-02 2.578e-02 -0.595
hp:speed
                                                                  0.5522
attack:dual typeTRUE
                                   -4.673e+00 2.263e+00 -2.065
                                                                  0.0393 *
                                   -1.145e+00 1.645e+00 -0.696
defense:dual_typeTRUE
                                                                  0.4865
hp:dual_typeTRUE
                                   -2.954e-01 2.103e+00 -0.140
                                                                  0.8883
                                   -1.211e+00 2.364e+00 -0.512
speed:dual_typeTRUE
                                                                  0.6088
                                    3.114e-04 2.575e-04 1.209
attack:defense:hp
                                                                 0.2269
                                    6.150e-04 3.243e-04
attack:defense:speed
                                                          1.896
                                                                  0.0583 .
                                    5.601e-04 3.237e-04
attack:hp:speed
                                                          1.730
                                                                  0.0840 .
                                    1.040e-04 3.763e-04
defense:hp:speed
                                                          0.276
                                                                  0.7823
                                    3.641e-02 2.533e-02
attack:defense:dual_typeTRUE
                                                          1.437
                                                                  0.1511
attack:hp:dual_typeTRUE
                                    2.736e-02 3.026e-02
                                                          0.904
                                                                  0.3663
                                   -3.543e-03 2.594e-02 -0.137
defense:hp:dual_typeTRUE
                                                                  0.8914
                                   6.003e-02 3.591e-02 1.671
attack:speed:dual_typeTRUE
                                                                  0.0951 .
defense:speed:dual_typeTRUE
                                   -9.959e-03 3.045e-02 -0.327
                                                                  0.7437
hp:speed:dual_typeTRUE
                                   -1.139e-02 3.673e-02 -0.310
                                                                  0.7565
attack:defense:hp:speed
                                   -4.839e-06 4.045e-06 -1.196
                                                                  0.2320
attack:defense:hp:dual typeTRUE
                                   -1.865e-04 3.375e-04 -0.552
                                                                 0.5808
attack:defense:speed:dual_typeTRUE
                                   -3.515e-04 3.985e-04 -0.882
                                                                  0.3781
                                   -3.609e-04 4.901e-04 -0.736
attack:hp:speed:dual_typeTRUE
                                                                  0.4617
defense:hp:speed:dual_typeTRUE
                                   3.258e-04 4.559e-04
                                                          0.715
                                                                  0.4750
attack:defense:hp:speed:dual_typeTRUE 1.376e-06 5.313e-06
                                                          0.259
                                                                  0.7957
```

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 54.88 on 768 degrees of freedom (1 observation deleted due to missingness)

Multiple R-squared: 0.5022, Adjusted R-squared: 0.4821 F-statistic: 24.99 on 31 and 768 DF, p-value: < 2.2e-16

Adjusted R-squared: 0.4821

Nested F-test

```
anova(add_capture_model3, interact_capture_model3)
```

Analysis of Variance Table

Model 1: capture_rate ~ attack + defense + hp + speed + dual_type

```
Model 2: capture_rate ~ attack * defense * hp * speed * dual_type
Res.Df RSS Df Sum of Sq F Pr(>F)
1 794 2528677
2 768 2313168 26 215508 2.752 7.923e-06 ***
---
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

The p-value when comparing the additive and interactive model for attack, speed, hp, defense, and dual_type is below 0.05 so it seems like there is some significance for going from the additive model to an interactive model. The difference in adjusted r-squared is 0.0298 which doesn't seem that significant.

Also when looking at the adjusted r-squared of the additive model when adding dual_type, when adding dual_type it is lower meaning it fits the model slightly less, so the extra complexity isn't needed since it doesn't help or model.

When looking at the adjusted r-squared between the interaction models shows that the dual_type model to the interaction model that shows that adding it improved the adjusted r-squared by 0.0039 which is really small so I argue it doesn't add to these 4 base stats based on adjusted r-squared.

2.586556

defense

2.293341

2.924339

attack

2.198357

```
additive_capture_model4 <- lm(capture_rate ~ attack + defense + hp + speed + base_total, defense + hp + speed + base
```

speed base_total

8.718589