

# TOPIC2 In Class Problems

# In-class Practice Problem 4

From the condominium problem:

- (a) Create a model with all interaction terms (based on the best model from last class)
- (b) Are interaction terms appropriate? Conduct a test to find out.

# ANSWER PROBLEM 4

Call:

```
lm(formula = listprice ~ livingarea + floors + baths, data = condo)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-11.796	-1.483	1.077	2.903	11.892

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	15.590	7.501	2.078	0.061888 .
livingarea	65.192	6.446	10.114	6.6e-07 ***
floors	-14.925	5.465	-2.731	0.019533 *
baths	28.381	5.715	4.966	0.000425 ***

---

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

Residual standard error: 6.622 on 11 degrees of freedom

Multiple R-squared: 0.9706, Adjusted R-squared: 0.9625

F-statistic: 120.9 on 3 and 11 DF, p-value: 1.059e-08

Model 1: listprice ~ livingarea + floors + baths +  
livingarea:baths +

livingarea:floors + floors:baths

Model 2: listprice ~ livingarea + floors + baths

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	8	349.67				
2	11	482.39	-3	-132.72	1.0121	0.4364

Call:

```
lm(formula = listprice ~ livingarea + floors + baths + livingarea:baths +  
livingarea:floors + floors:baths, data = condo)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-10.2701	-2.0116	-0.4466	3.8389	6.2406

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	40.502	77.021	0.526	0.613
livingarea	27.339	44.761	0.611	0.558
floors	-19.319	109.400	-0.177	0.864
baths	28.625	24.946	1.147	0.284
livingarea:baths	10.209	31.157	0.328	0.752
livingarea:floors	13.143	25.315	0.519	0.618
floors:baths	-8.062	37.341	-0.216	0.834

Residual standard error: 6.611 on 8 degrees of freedom

Multiple R-squared: 0.9787, Adjusted R-squared: 0.9626

F-statistic: 61.13 on 6 and 8 DF, p-value: 3.008e-06

Analysis of Variance Table

Sales.csv

# In-class Practice Problem 5

Data on last year's sale (Y in 100,000s dollars) in 40 sales districts are given in the sales.csv file. This file also contains:

- promotional expenditures ( $X_1$ : in 1,000s dollars),
- the number of active accounts ( $X_2$ ),
- the number of competing brands ( $X_3$ ) and
- the district potential ( $X_4$ , coded) for each of the district (OMIT THIS VARIABLE FOR NOW)

1. Find the best fit additive to predict sales using some or all of the variables  $X_1, X_2, X_3$  only.
2. Find the best fit model with interaction terms (if needed) using some or all of the variables  $X_1, X_2, X_3$
3. Which model would you choose? Explain.
4. Once you obtain the best fit model, interpret the regression coefficient for  $X_3$  (Hint: it will interact with another variable).

$X_3 \rightarrow$  involve an interaction

# ANSWER PROBLEM 5 (1 of 3)

## PART 1

Call:

```
lm(formula = Y ~ X1 + X2 + X3, data = sale)
```

Residuals:

Min	1Q	Median	3Q	Max
-106.803	-6.726	-1.967	7.072	81.964

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	162.2269	31.0376	5.227	7.50e-06 ***
X1	2.0192	2.5763	0.784	0.438
X2	3.4568	0.3426	10.088	4.91e-12 ***
X3	-19.4589	1.8054	-10.778	8.08e-13 ***

---

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

Residual standard error: 25.35 on 36 degrees of freedom

Multiple R-squared: 0.9175, Adjusted R-squared: 0.9106

F-statistic: 133.4 on 3 and 36 DF, p-value: < 2.2e-16

Call:

```
lm(formula = Y ~ X2 + X3, data = sale)
```

Residuals:

Min	1Q	Median	3Q	Max
-109.096	-5.888	-3.440	8.780	83.982

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	172.4595	28.0109	6.157	3.85e-07 ***
X2	3.5011	0.3362	10.414	1.50e-12 ***
X3	-19.7308	1.7625	-11.195	1.94e-13 ***

---

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

Residual standard error: 25.22 on 37 degrees of freedom

Multiple R-squared: 0.9161, Adjusted R-squared: 0.9115

F-statistic: 201.9 on 2 and 37 DF, p-value: < 2.2e-16

Analysis of Variance Table

Model 1: Y ~ X2 + X3

Model 2: Y ~ X1 + X2 + X3

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	37	23532				
2	36	23137	1	394.79	0.6148	0.4383

# ANSWER PROBLEM 5 (2 of 3)

## PART 2

Call:  
lm(formula = Y ~ X2 + X3 + X2 : X3, data = sale)

Residuals:

Min	1Q	Median	3Q	Max
-98.788	-6.804	-1.861	6.225	58.055

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	19.3191	62.5599	0.309	0.7592
X2	6.0809	1.0084	6.030	6.33e-07 ***
X3	-2.9261	6.4576	-0.453	0.6532
X2:X3	-0.2903	0.1079	-2.689	0.0108 *

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1  
Residual standard error: 23.33 on 36 degrees of freedom  
Multiple R-squared: 0.9301, Adjusted R-squared: 0.9243  
F-statistic: 159.7 on 3 and 36 DF, p-value: < 2.2e-16

Call:  
lm(formula = Y ~ (X1 + X2 + X3)^2, data = sale)

Residuals:

Min	1Q	Median	3Q	Max
-93.253	-9.208	0.852	6.606	51.455

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-22.0421	91.7901	-0.240	0.81171
X1	11.9200	14.7995	0.805	0.42633
X2	4.8325	1.6708	2.892	0.00672 **
X3	7.0945	7.8501	0.904	0.37268
X1:X2	0.1193	0.2169	0.550	0.58607
X1:X3	-2.1302	0.9540	-2.233	0.03246 *
X2:X3	-0.2495	0.1168	-2.136	0.04021 *

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1  
Residual standard error: 22.59 on 33 degrees of freedom  
Multiple R-squared: 0.94, Adjusted R-squared: 0.929  
F-statistic: 86.09 on 6 and 33 DF, p-value: < 2.2e-16

Call:  
lm(formula = Y ~ X1 + X2 + X3 + X2 : X3 + X1 : X3, data = sale)

Residuals:

Min	1Q	Median	3Q	Max
-93.180	-9.362	0.929	7.712	53.205

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-55.0832	68.6781	-0.802	0.4281
X1	18.4248	8.8028	2.093	0.0439 *
X2	5.5102	1.1166	4.935	2.09e-05 ***
X3	6.9378	7.7640	0.894	0.3778
X2:X3	-0.2524	0.1155	-2.185	0.0358 *
X1:X3	-2.1387	0.9441	-2.265	0.0300 *

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

Residual standard error: 22.35 on 34 degrees of freedom  
Multiple R-squared: 0.9394, Adjusted R-squared: 0.9305  
F-statistic: 105.4 on 5 and 34 DF, p-value: < 2.2e-16

# ANSWER PROBLEM 5 (3 of 3)

1. Which model would you choose? Explain.
2. Once you obtain the best fit model, interpret the regression coefficient for  $X_3$  (Hint: it will interact with another variable).

```
Call:
lm(formula = Y ~ X1 + X2 + X3 + X2 : X3 + X1 : X3, data = sale)

Residuals:
    Min       1Q   Median       3Q      Max
-93.180  -9.362   0.929   7.712  53.205

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)  -55.0832     68.6781  -0.802   0.4281
X1             18.4248     8.8028   2.093   0.0439 *
X2              5.5102     1.1166   4.935 2.09e-05 ***
X3              6.9378     7.7640   0.894   0.3778
X2:X3         -0.2524     0.1155  -2.185   0.0358 *
X1:X3         -2.1387     0.9441  -2.265   0.0300 *
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 22.35 on 34 degrees of freedom
Multiple R-squared:  0.9394,    Adjusted R-squared:  0.9305
F-statistic: 105.4 on 5 and 34 DF,  p-value: < 2.2e-16
```

# Inclass Practice Problem 6

Suppose that we wish to investigate differences in credit card balance between marital status. Based on the Married variable, we can create a dummy variable which 0 is NO and 1 is Yes.

- (a) Create a simple linear regression model to predict the credit card balance by using the Married variable.
- (b) How much is the average credit card debt for an unmarried person.
- (c) What is the difference in debt between a married and single person.

**Ignore the individual t-test output**



# ANSWER PROBLEM 6

Call:

```
lm(formula = Balance ~ factor(Married), data = credit)
```

Residuals:

Min	1Q	Median	3Q	Max
-523.29	-451.03	-60.12	345.06	1481.06

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	523.290	36.974	14.153	<2e-16 ***
factor(Married)Yes	-5.347	47.244	-0.113	0.91

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

Residual standard error: 460.3 on 398 degrees of freedom

Multiple R-squared: 3.219e-05, Adjusted R-squared: -0.00248

F-statistic: 0.01281 on 1 and 398 DF, p-value: 0.9099

number	Income	Limit	Rating	Cards	Age	Education	Gender	Student	Married	
1	1	14.891	3606	283	2	34	11	Male	No	Yes
2	2	106.025	6645	483	3	82	15	Female	Yes	Yes
3	3	104.593	7075	514	4	71	11	Male	No	No
4	4	148.924	9504	681	3	36	11	Female	No	No
5	5	55.882	4897	357	2	68	16	Male	No	Yes
6	6	80.180	8047	569	4	77	10	Male	No	No

Ethnicity	Balance
1 Caucasian	333
2 Asian	903
3 Asian	580
4 Asian	964
5 Caucasian	331
6 Caucasian	1151

$$\hat{y} = \hat{\beta}_0 + \hat{\beta}_1 x_1$$

$\hat{y} = 523.29 - 5.347 [\text{Married}]$

$\hat{\beta}_0 = 523.29 \rightarrow$  Average debt for unmarried. ( $x_1 = 0$ )

$\hat{\beta}_1 = -5.35 \rightarrow$  Difference (between levels)

$\hat{\beta}_0 + \hat{\beta}_1 = 523.29 + (-5.35) = 517.94$  average debt for married.

# In-class Practice Problem 7

There is always a certain curiosity and controversy surrounding professors' salaries and whether they are overpaid or not paid enough. A university would like to study the effects of ranks and departments on salaries. 30 observations were randomly chosen from 3 different departments. The data are provided in the salary.csv data file. Dept= Department (1=Family Studies, 2=Biology, 3=Business)

Instead of the rank variable, practice how to interpret the dept variable.

STATE THE AVERAGE SALARY OF  
A PERSON FROM EACH DEPT

# ANSWER PROBLEM 7

Call:

```
lm(formula = salary ~ factor(dept), data = salary)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-12.250	-6.838	-3.925	4.662	30.000

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	42.250	2.788	15.154	1.01e-14 ***
factor(dept)2	7.750	4.408	1.758	0.09008 .
factor(dept)3	12.350	4.135	2.986	0.00594 **

---

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

Residual standard error: 9.658 on 27 degrees of freedom

Multiple R-squared: 0.2543, Adjusted R-squared: 0.199

F-statistic: 4.603 on 2 and 27 DF, p-value: 0.01905

$$\hat{\text{salary}} = \hat{\beta}_0 + \hat{\beta}_1 x_1 + \hat{\beta}_2 x_2$$

$\hat{\text{salary}} = \begin{cases} \hat{\beta}_0 & \text{FAMILY STUDIOS } (0,0) \\ \hat{\beta}_0 + \hat{\beta}_1 & \text{BIOLOGY } (1,0) \\ \hat{\beta}_0 + \hat{\beta}_2 & \text{BUSINESS } (0,1) \end{cases}$

Average

FAMILY STUDIOS

$$\$ \hat{\beta}_0 \times 1000 = \$42,250$$

BIOLOGY

$$\$ (\hat{\beta}_0 + \hat{\beta}_1) \times 1000 = \$42,250 + 7,750$$

BUSINESS

$$\$ (\hat{\beta}_0 + \hat{\beta}_2) \times 1000 = 42,250 + 12,350$$

# Inclass Practice Problem 8

From the credit card example, use the `lm()` function to perform the best fit model. How would you interpret the regression coefficients (if possible)? Would you recommend this model for predictive purpose?

## STEPS

1. Build additive model
2. Determine significant predictors
3. Build interaction model with significant predictors
4. Remove non-significant interactions

(a) Build full additive model with only significant predictors

(b) Build interacting model with predictors from (a)

(c) Remove non-significant interactions and rerun model

5. Rerun model to ensure all predictors significant
6. Iterate at step 5 until done.

# ANSWER PROBLEM 8

- (a) Build full additive model with only significant predictors
- (b) Build interacting model with predictors from (a)
- (c) Remove non-significant interactions and rerun model
- (d) Interpret the final model

Call:

```
lm(formula = Balance ~ Income + Limit + Rating + Cards + Age +  
    Education + factor(Gender) + factor(Ethnicity) + factor(Married) +  
    factor(Student), data = credit)
```

Residuals:

Min	1Q	Median	3Q	Max
-161.64	-77.70	-13.49	53.98	318.20

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	-479.20787	35.77394	-13.395	< 2e-16	***
Income	-7.80310	0.23423	-33.314	< 2e-16	***
Limit	0.19091	0.03278	5.824	1.21e-08	***
Rating	1.13653	0.49089	2.315	0.0211	*
Cards	17.72448	4.34103	4.083	5.40e-05	***
Age	-0.61391	0.29399	-2.088	0.0374	*
Education	-1.09886	1.59795	-0.688	0.4921	
factor(Gender)Female	-10.65325	9.91400	-1.075	0.2832	
factor(Ethnicity)Asian	16.80418	14.11906	1.190	0.2347	
factor(Ethnicity)Caucasian	10.10703	12.20992	0.828	0.4083	
factor(Married)Yes	-8.53390	10.36287	-0.824	0.4107	
factor(Student)Yes	425.74736	16.72258	25.459	< 2e-16	***

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

Residual standard error: 98.79 on 388 degrees of freedom

Multiple R-squared: 0.9551, Adjusted R-squared: 0.9538

F-statistic: 750.3 on 11 and 388 DF, p-value: < 2.2e-16

# ANSWER PROBLEM 8

- (a) Build full additive model with only significant predictors
- (b) Build interacting model with predictors from (a)
- (c) Remove non-significant interactions and rerun model
- (d) Interpret the final model

Call:

```
lm(formula = Balance ~ (Income + Limit + Rating + Cards + Age +  
  factor(Student))^2, data = credit)
```

Residuals:

Min	1Q	Median	3Q	Max
-166.579	-40.014	8.191	38.844	163.054

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-2.923e+02	4.966e+01	-5.886	8.72e-09 ***
Income	-1.907e+00	8.011e-01	-2.381	0.01777 *
Limit	3.230e-03	8.354e-02	0.039	0.96918
Rating	1.446e+00	1.252e+00	1.154	0.24912
Cards	8.495e+00	1.426e+01	0.596	0.55182
Age	9.420e-01	7.315e-01	1.288	0.19862
factor(Student)Yes	1.909e+02	6.589e+01	2.898	0.00398 **
Income:Limit	6.667e-04	5.931e-04	1.124	0.26168
Income:Rating	-2.708e-02	8.703e-03	-3.112	0.00200 **
Income:Cards	-1.755e-01	1.247e-01	-1.407	0.16021
Income:Age	1.878e-02	8.833e-03	2.126	0.03414 *
Income:factor(Student)Yes	-1.565e+00	4.769e-01	-3.282	0.00113 **
Limit:Rating	3.420e-04	1.751e-05	19.536	< 2e-16 ***
Limit:Cards	3.130e-03	1.168e-02	0.268	0.78883
Limit:Age	8.277e-04	1.281e-03	0.646	0.51860
Limit:factor(Student)Yes	2.075e-01	6.806e-02	3.048	0.00247 **
Rating:Cards	-4.870e-03	1.734e-01	-0.028	0.97761
Rating:Age	-1.869e-02	1.919e-02	-0.974	0.33075
Rating:factor(Student)Yes	-1.966e+00	1.019e+00	-1.929	0.05447 .
Cards:Age	3.773e-02	1.748e-01	0.216	0.82920
Cards:factor(Student)Yes	1.073e+01	9.452e+00	1.136	0.25678
Age:factor(Student)Yes	2.499e-01	7.669e-01	0.326	0.74475

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

Residual standard error: 62.94 on 378 degrees of freedom

Multiple R-squared: 0.9822, Adjusted R-squared: 0.9813

F-statistic: 995.8 on 21 and 378 DF, p-value: < 2.2e-16

# ANSWER PROBLEM 8

- (a) Build full additive model with only significant predictors
- (b) Build interacting model with predictors from (a)
- (c) Remove non-significant interactions and rerun model
- (d) Interpret the final model

Call:

```
lm(formula = Balance ~ Income + Limit + Rating + Cards + Age +  
    factor(Student) + Income * Age + Income * Rating + Income *  
    factor(Student) + Limit * Rating + Limit * factor(Student) +  
    Rating * factor(Student), data = credit)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-218.008	-42.145	7.003	39.734	147.616

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-2.058e+02	2.531e+01	-8.130	5.86e-15 ***
Income	-1.684e+00	5.693e-01	-2.958	0.003291 **
Limit	1.008e-01	2.251e-02	4.478	9.91e-06 ***
Rating	-2.018e-01	3.334e-01	-0.605	0.545365
Cards	1.810e+01	2.792e+00	6.482	2.76e-10 ***
Age	-6.310e-01	3.107e-01	-2.031	0.042970 *
factor(Student)Yes	1.957e+02	4.276e+01	4.578	6.34e-06 ***
Income:Age	-3.194e-03	5.149e-03	-0.620	0.535339
Income:Rating	-1.687e-02	1.198e-03	-14.078	< 2e-16 ***
Income:factor(Student)Yes	-1.703e+00	4.500e-01	-3.784	0.000179 ***
Limit:Rating	3.367e-04	1.717e-05	19.605	< 2e-16 ***
Limit:factor(Student)Yes	1.498e-01	5.999e-02	2.497	0.012955 *
Rating:factor(Student)Yes	-1.067e+00	8.910e-01	-1.198	0.231808

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

Residual standard error: 63.6 on 387 degrees of freedom

Multiple R-squared: 0.9814, Adjusted R-squared: 0.9809

F-statistic: 1705 on 12 and 387 DF, p-value: < 2.2e-16

# ANSWER PROBLEM 8

- (a) Build full additive model with only significant predictors
- (b) Build interacting model with predictors from (a)
- (c) Remove non-significant interactions and rerun model
- (d) Interpret the final model

Call:

```
lm(formula = Balance ~ Income + Limit + Rating + Cards + Age +  
    factor(Student) + Income * Rating + Income * factor(Student) +  
    Limit * Rating + Limit * factor(Student), data = credit)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-231.817	-41.097	7.283	38.913	153.038

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-1.945e+02	2.160e+01	-9.006	< 2e-16 ***
Income	-1.837e+00	5.235e-01	-3.508	0.000504 ***
Limit	1.079e-01	2.158e-02	5.000	8.70e-07 ***
Rating	-3.121e-01	3.200e-01	-0.976	0.329914
Cards	1.832e+01	2.786e+00	6.575	1.57e-10 ***
Age	-7.660e-01	1.886e-01	-4.063	5.87e-05 ***
factor(Student)Yes	1.555e+02	2.634e+01	5.905	7.68e-09 ***
Income:Rating	-1.694e-02	1.187e-03	-14.272	< 2e-16 ***
Income:factor(Student)Yes	-1.784e+00	4.460e-01	-4.001	7.55e-05 ***
Limit:Rating	3.373e-04	1.711e-05	19.710	< 2e-16 ***
Limit:factor(Student)Yes	7.868e-02	7.666e-03	10.264	< 2e-16 ***

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

keep it in  
due to  
hierarchical  
principle

Residual standard error: 63.6 on 389 degrees of freedom

Multiple R-squared: 0.9813, Adjusted R-squared: 0.9809

F-statistic: 2046 on 10 and 389 DF, p-value: < 2.2e-16



# ANSWER PROBLEM 8

- (a) Build full additive model with only significant predictors
- (b) Build interacting model with predictors from (a)
- (c) Remove non-significant interactions and rerun model
- (d) Interpret the final model

Call:

```
lm(formula = Balance ~ Income + Limit + Rating + Cards + Age +  
    factor(Student) + Income * Rating + Income * factor(Student) +  
    Limit * Rating + Limit * factor(Student), data = credit)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-231.817	-41.097	7.283	38.913	153.038

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-1.945e+02	2.160e+01	-9.006	< 2e-16 ***
Income	-1.837e+00	5.235e-01	-3.508	0.000504 ***
Limit	1.079e-01	2.158e-02	5.000	8.70e-07 ***
Rating	-3.121e-01	3.200e-01	-0.976	0.329914
Cards	1.832e+01	2.786e+00	6.575	1.57e-10 ***
Age	-7.660e-01	1.886e-01	-4.063	5.87e-05 ***
factor(Student)Yes	1.555e+02	2.634e+01	5.905	7.68e-09 ***
Income:Rating	-1.694e-02	1.187e-03	-14.272	< 2e-16 ***
Income:factor(Student)Yes	-1.784e+00	4.460e-01	-4.001	7.55e-05 ***
Limit:Rating	3.373e-04	1.711e-05	19.710	< 2e-16 ***
Limit:factor(Student)Yes	7.868e-02	7.666e-03	10.264	< 2e-16 ***

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

Residual standard error: 63.6 on 389 degrees of freedom

Multiple R-squared: 0.9813, Adjusted R-squared: 0.9809

F-statistic: 2046 on 10 and 389 DF, p-value: < 2.2e-16

## TRY THIS ON YOUR OWN

$$\begin{aligned}\hat{y} &= b_0 + b_1 \text{Income} + b_2 \text{Limit} + b_3 \text{Rating} + b_4 \text{Cards} + b_5 \text{Age} + b_6 \text{Student} \\ &+ b_7 \text{Income} * \text{Rating} + b_8 \text{Income} * \text{Student} + b_9 \text{Limit} * \text{Rating} + b_{10} \text{Limit} * \text{Student} \\ &\hat{y} = -0.01945 - 1.837 \text{Income} + 0.1079 \text{Limit} - 0.3121 \text{Rating} + 10.832 \text{Cards} \\ &\quad - 0.7660 \text{Age} + 155.5 \text{Student} - 0.01694 \text{Income} * \text{Rating} \\ &\quad - 1.784 \text{Income} * \text{Student} + 0.0003373 \text{Limit} * \text{Rating} + 0.07868 \text{Limit} * \text{Student}\end{aligned}$$

# ANSWER PROBLEM 8

- (a) Build full additive model with only significant predictors
- (b) Build interacting model with predictors from (a)
- (c) Remove non-significant interactions and rerun model
- (d) Interpret the final model

Call:

```
lm(formula = Balance ~ Income + Limit + Rating + Cards + Age +  
    factor(Student) + Income * Rating + Income * factor(Student) +  
    Limit * Rating + Limit * factor(Student), data = credit)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-231.817	-41.097	7.283	38.913	153.038

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-1.945e+02	2.160e+01	-9.006	< 2e-16 ***
Income	-1.837e+00	5.235e-01	-3.508	0.000504 ***
Limit	1.079e-01	2.158e-02	5.000	8.70e-07 ***
Rating	-3.121e-01	3.200e-01	-0.976	0.329914
Cards	1.832e+01	2.786e+00	6.575	1.57e-10 ***
Age	-7.660e-01	1.886e-01	-4.063	5.87e-05 ***
factor(Student)Yes	1.555e+02	2.634e+01	5.905	7.68e-09 ***
Income:Rating	-1.694e-02	1.187e-03	-14.272	< 2e-16 ***
Income:factor(Student)Yes	-1.784e+00	4.460e-01	-4.001	7.55e-05 ***
Limit:Rating	3.373e-04	1.711e-05	19.710	< 2e-16 ***
Limit:factor(Student)Yes	7.868e-02	7.666e-03	10.264	< 2e-16 ***

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

Residual standard error: 63.6 on 389 degrees of freedom

Multiple R-squared: 0.9813, Adjusted R-squared: 0.9809

F-statistic: 2046 on 10 and 389 DF, p-value: < 2.2e-16

## TRY THIS ON YOUR OWN

$$\hat{y} = b_0 + b_1 \text{Income} + b_2 \text{Limit} + b_3 \text{Rating} + b_4 \text{Cards} + b_5 \text{Age} + b_6 \text{Student} \\ + b_7 \text{Income} * \text{Rating} + b_8 \text{Income} * \text{Student} + b_9 \text{Limit} * \text{Rating} + b_{10} \text{Limit} * \text{Student}$$

$$\hat{y} = -0.01945 - 1.837 \text{Income} + 0.1079 \text{Limit} - 0.3121 \text{Rating} + 10.832 \text{Cards} \\ - 0.7660 \text{Age} + 155.5 \text{Student} - 0.01694 \text{Income} * \text{Rating} \\ - 1.784 \text{Income} * \text{Student} + 0.0003373 \text{Limit} * \text{Rating} + 0.07868 \text{Limit} * \text{Student}$$

# ANSWER PROBLEM 8

- (a) Build full additive model with only significant predictors
- (b) Build interacting model with predictors from (a)
- (c) Remove non-significant interactions and rerun model
- (d) Interpret the final model

This is our final model  $\rightarrow$  let's interpret it.

$$\hat{y} = -0.01945 - 1.837\text{Income} + 0.1079\text{Limit} - 0.3121\text{Rating} + 10.832\text{Cards} \\ - 0.7660\text{Age} + 155.5\text{Student} - 0.01694\text{Income} * \text{Rating} \\ - 1.784\text{Income} * \text{Student} + 0.0003373\text{Limit} * \text{Rating} + 0.07868\text{Limit} * \text{Student}$$

student = 0

WHAT IS THE EFFECT ON INCOME IF NOT A STUDENT

$\downarrow = 0$

$$-1.837[\text{income}] - 0.01694[\text{income}][\text{rating}] - 1.784[\text{income}][\text{student}]$$

$$(-1.837 - 0.01694[\text{rating}])[\text{income}] - 0$$

Factor out  
0 to make  
easier to  
explain

$$(-1.837 + 0.01694[\text{rating}])$$

IS THIS AN INCREASE OR DECREASE?

$\rightarrow$  decrease

# In-class Practice Problem 9

Suppose you wanted to model the quality,  $y$ , of a product as a function of the pressure pounds per square inch (psi), at which it is produced. Four inspectors independently assign a quality score between 0 and 100 to each product, and then the quality,  $y$ , is calculated by averaging the four scores. An experiment is conducted by varying temperature in F. The data are provided in **PRODQUAL.csv** file

Fit a higher-order model to the data and sketch the scatterplot.

Which order would you select?

QUALITY ~ PRESSURE

Find the highest order model that we could use.

# ANSWER PROBLEM 9

```
Call:
lm(formula = QUALITY ~ PRESSURE + I(PRESSURE^2), data = quality)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-12.136	-6.234	-2.852	7.660	16.410

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-3.791e+03	2.857e+02	-13.27	<2e-16 ***
PRESSURE	1.423e+02	1.039e+01	13.70	<2e-16 ***
I(PRESSURE^2)	-1.307e+00	9.418e-02	-13.88	<2e-16 ***

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

Residual standard error: 7.956 on 73 degrees of freedom  
Multiple R-squared: 0.7622, Adjusted R-squared: 0.7557  
F-statistic: 117 on 2 and 73 DF, p-value: < 2.2e-16

```
Call:
lm(formula = QUALITY ~ PRESSURE + I(PRESSURE^2) + I(PRESSURE^3),
    data = quality)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-12.430	-5.536	-0.779	5.710	15.170

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-3.083e+04	6.089e+03	-5.064	3.04e-06 ***
PRESSURE	1.623e+03	3.332e+02	4.871	6.38e-06 ***
I(PRESSURE^2)	-2.827e+01	6.065e+00	-4.661	1.41e-05 ***
I(PRESSURE^3)	1.633e-01	3.672e-02	4.446	3.12e-05 ***

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

Residual standard error: 7.097 on 72 degrees of freedom  
Multiple R-squared: 0.8134, Adjusted R-squared: 0.8056  
F-statistic: 104.6 on 3 and 72 DF, p-value: < 2.2e-16

```
Call:
lm(formula = QUALITY ~ PRESSURE + I(PRESSURE^2) + I(PRESSURE^3) +
    I(PRESSURE^4), data = quality)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-15.3715	-4.4458	-0.7475	3.9742	13.2232

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	4.958e+05	1.208e+05	4.106	0.000106 ***
PRESSURE	-3.669e+04	8.780e+03	-4.178	8.24e-05 ***
I(PRESSURE^2)	1.015e+03	2.391e+02	4.246	6.48e-05 ***
I(PRESSURE^3)	-1.245e+01	2.890e+00	-4.309	5.18e-05 ***
I(PRESSURE^4)	5.710e-02	1.308e-02	4.366	4.22e-05 ***

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

Residual standard error: 6.345 on 71 degrees of freedom  
Multiple R-squared: 0.8529, Adjusted R-squared: 0.8446  
F-statistic: 102.9 on 4 and 71 DF, p-value: < 2.2e-16

```
ggplot(data=quality) +
  aes(x=PRESSURE, y=QUALITY) +
  geom_point(color='red') +
  geom_smooth()
```

