MATH 212 - Final Project

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Data Set

From https://fred.stlouisfed.org/, I retrieved a data set with various important economic indicators over the period from Q2 2011 through Q1 2020 (immediately before the COVID-induced recession).

econ_indicators

```
## # A tibble: 36 x 6
                                      GDP MEDIAN_WEELKY_WAGES
##
      observation_date
                            UNRATE
                                                                 DJCA HOUST
##
      <dttm>
                                    <dbl>
                                                          <dbl> <dbl>
                             <dbl>
                                                                      <dbl>
##
    1 2011-04-01 00:00:00
                               9.1 15496.
                                                            754 4271.
                                                                        1723
                                   15592.
    2 2011-07-01 00:00:00
                                                            760 4008.
##
                                                                        1858
    3 2011-10-01 00:00:00
                               8.5 15796.
                                                            760 4068.
                                                                        2015
                                                                        2122
##
    4 2012-01-01 00:00:00
                               8.2 16020.
                                                            764 4377.
##
    5 2012-04-01 00:00:00
                               8.2 16152.
                                                            772 4365.
                                                                        2218
    6 2012-07-01 00:00:00
                                                            766 4435.
##
                               7.8 16257.
                                                                        2341
      2012-10-01 00:00:00
                               7.9 16359.
                                                            771 4413.
                                                                        2724
##
##
    8 2013-01-01 00:00:00
                               7.5 16570.
                                                            768 4792.
                                                                        2860
    9 2013-04-01 00:00:00
                               7.5 16638.
                                                            777 5091.
                                                                        2604
## 10 2013-07-01 00:00:00
                               7.2 16849.
                                                            779 5181.
                                                                        2647
## # ... with 26 more rows
```

- 1. There are 5 variables in my data set.
- Unemployment rate (UNRATE) From FRED: "The unemployment rate represents the number of unemployed as a percentage of the labor force. Labor force data are restricted to people 16 years of age and older, who currently reside in 1 of the 50 states or the District of Columbia, who do not reside in institutions (e.g., penal and mental facilities, homes for the aged), and who are not on active duty in the Armed Forces. This rate is also defined as the U-3 measure of labor underutilization." Unit is percent (seasonally adjusted).
- Gross domestic product (GDP) From FRED: "Gross domestic product (GDP), the featured measure of U.S. output, is the market value of the goods and services produced by labor and property located in the United States." Unit is billions of dollars, seasonally adjusted annual rate.
- Employed full time: median usual weekly nominal earnings, 16 years and over (MEDIAN_EARNINGS) from FRED: "Data measure usual weekly earnings of wage and salary workers. Wage and salary workers are workers who receive wages, salaries, commissions, tips, payment in kind, or piece rates. The group includes employees in both the private and public sectors but, for the purposes of the earnings series, it excludes all self-employed persons, both those with incorporated businesses and those with unincorporated businesses. Usual weekly earnings represent earnings before taxes and other deductions and include any overtime pay, commissions, or tips usually received (at the main job in the case of multiple jobholders). Prior to 1994, respondents were asked how much they usually earned per week. Since January 1994, respondents have been asked to identify the easiest way for them to report earnings (hourly, weekly, biweekly, twice monthly, monthly, annually, or other) and how much they usually earn in the reported time period. Earnings reported on a basis other than weekly are converted to a weekly

- equivalent. The term "usual" is determined by each respondent's own understanding of the term. If the respondent asks for a definition of "usual," interviewers are instructed to define the term as more than half the weeks worked during the past 4 or 5 months." Unit is dollars, seasonally adjusted.
- Dow Jones Composite Average (DJCA) From FRED: "The Dow Jones Composite Average is combination of all three major Dow Jones Averages (Industrial, Utility, and Transportation). Since the Composite Average is made up of this select group of prominent stocks, Dow Jones refers to it as a blue chip microcosm of the US stock market." Unit is index, not seasonally adjusted.
- Housing Starts (HOUST) From FRED: "As provided by the Census, start occurs when excavation begins for the footings or foundation of a building. All housing units in a multifamily building are defined as being started when this excavation begins. Beginning with data for September 1992, estimates of housing starts include units in structures being totally rebuilt on an existing foundation."
- 2. My response variable in median weekly wages, while my predictor variables are unemployment, GDP, DJCA, and housing starts.
- 3. My research question is: can we linearly predict median weekly wages given some combination of unemployment, GDP, DJCA, and housing starts? I think this is an interesting question because median weekly wages provides a stronger indication of the economic conditions for actual working people in America. GDP, while important, doesn't tell us precisely if people are thriving or not. Hence, I want to see how median wages can be predicted from these other important economic indicators.

Correlation Analysis

1. Create correlation plot.

```
MEDIAN_WEEKLY_WAGES <- econ_indicators$MEDIAN_WEELKY_WAGES

UNRATE <- econ_indicators$UNRATE

GDP <- econ_indicators$GDP

DJCA <- econ_indicators$DJCA

HOUST <- econ_indicators$HOUST

DF_indicators <- data.frame(MEDIAN_WEEKLY_WAGES, UNRATE, GDP, DJCA, HOUST)

r <- cor(DF_indicators)

ggcorrplot(r, title="Economic Indicators (data from FRED)", lab=TRUE)
```

Economic Indicators (data from FRED)



- 2. The matrix indicates very high correlation between MEDIAN_WEEKLY_WAGES and both GDP and DJCA (0.98 and 0.96). Either of these predictor variables seem to be good candidates for being the first to enter our stepwise model.
- 3. All of our predictor values are correlated with our response variable (MEDIAN_WEEKLY_WAGES). UNRATE is negatively correlated, while HOUST, GDP, and DJCA are all positively correlated with our response variable.
- 4. Given how closely related some of these indicators are, we, unsurprisingly, do see multicolinearity in the dataset. For example, we see that GDP and DJCA are highly correlated (0.99).

Variable Selection

1. Regress MEDIAN_WEEKLY_WAGES on each predictor variable and compare outputs. Use alpha_to_enter = 0.15 and alpha_to_remove = 0.15.

```
model1 <- lm(MEDIAN_WEEKLY_WAGES ~ DJCA, data=DF_indicators)</pre>
summary(model1)
##
## Call:
## lm(formula = MEDIAN WEEKLY WAGES ~ DJCA, data = DF indicators)
##
## Residuals:
##
       Min
                 1Q
                     Median
                                  3Q
                                         Max
##
  -29.148
            -8.956
                      5.366
                               9.865
                                     41.847
##
## Coefficients:
```

```
Estimate Std. Error t value Pr(>|t|)
## (Intercept) 6.068e+02 1.083e+01 56.04
                                            <2e-16 ***
                                             <2e-16 ***
## DJCA
              3.433e-02 1.639e-03 20.95
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 15.41 on 34 degrees of freedom
## Multiple R-squared: 0.9281, Adjusted R-squared: 0.926
## F-statistic: 439 on 1 and 34 DF, p-value: < 2.2e-16
model2 <- lm(MEDIAN_WEEKLY_WAGES ~ GDP, data=DF_indicators)</pre>
summary(model2)
##
## Call:
## lm(formula = MEDIAN_WEEKLY_WAGES ~ GDP, data = DF_indicators)
## Residuals:
      Min
               1Q Median
                               3Q
## -17.242 -5.412 -1.048
                            6.200 32.190
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 2.930e+02 1.731e+01 16.92 <2e-16 ***
## GDP
              2.893e-02 9.327e-04 31.02 <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 10.62 on 34 degrees of freedom
## Multiple R-squared: 0.9659, Adjusted R-squared: 0.9649
## F-statistic: 962.4 on 1 and 34 DF, p-value: < 2.2e-16
model3 <- lm(MEDIAN_WEEKLY_WAGES ~ UNRATE, data=DF_indicators)</pre>
summary(model3)
##
## Call:
## lm(formula = MEDIAN_WEEKLY_WAGES ~ UNRATE, data = DF_indicators)
## Residuals:
##
               1Q Median
                               3Q
      Min
                                      Max
## -37.204 -19.608 -3.611 15.235 85.521
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 990.160
                        14.999 66.01 < 2e-16 ***
## UNRATE
               -28.791
                            2.536 -11.36 4.09e-13 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 26.25 on 34 degrees of freedom
## Multiple R-squared: 0.7913, Adjusted R-squared: 0.7852
## F-statistic: 128.9 on 1 and 34 DF, p-value: 4.087e-13
model4 <- lm(MEDIAN_WEEKLY_WAGES ~ HOUST, data=DF_indicators)</pre>
summary(model4)
```

```
##
## Call:
## lm(formula = MEDIAN_WEEKLY_WAGES ~ HOUST, data = DF_indicators)
## Residuals:
##
     Min
             1Q Median
                            3Q
                                 Max
  -44.95 -21.14 -6.13 21.36
##
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
  (Intercept) 5.906e+02 2.185e+01
                                     27.03 < 2e-16 ***
              7.409e-02 6.695e-03
                                     11.07 8.19e-13 ***
## HOUST
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 26.79 on 34 degrees of freedom
## Multiple R-squared: 0.7827, Adjusted R-squared: 0.7763
## F-statistic: 122.5 on 1 and 34 DF, p-value: 8.189e-13
```

Since the p-value for each of the predictor variables' slopes is less than our alpha_to_enter, each is a candidate to be entered into the stepwise model. We choose the one with the lowest p-value. Both the MEDIAN_WEEKLY_WAGES ~ GDP and MEDIAN_WEEKLY_WAGES ~ DJCA have slopes with p-value < 2e-16, but we notice that the t-value for the slope in the GDP model is higher, indicating greater significance. Thus, we'll choose GDP as the first variable for our model.

2. Fit each of the two-predictor models that include GDP as a predictor.

##

```
model1 <- lm(MEDIAN_WEEKLY_WAGES ~ GDP + DJCA, data=DF_indicators)</pre>
summary(model1)
##
## Call:
## lm(formula = MEDIAN_WEEKLY_WAGES ~ GDP + DJCA, data = DF_indicators)
## Residuals:
##
                  1Q
                                     3Q
        Min
                       Median
                                             Max
  -17.9655 -6.9021 -0.1618
                                6.3199
                                        30.0933
##
## Coefficients:
                 Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 201.530534
                           63.562538
                                       3.171 0.00328 **
## GDP
                 0.037562
                            0.005851
                                       6.419 2.82e-07 ***
## DJCA
                -0.010574
                            0.007083
                                      -1.493 0.14495
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 10.43 on 33 degrees of freedom
## Multiple R-squared: 0.968, Adjusted R-squared: 0.9661
## F-statistic: 499.7 on 2 and 33 DF, p-value: < 2.2e-16
model2 <- lm(MEDIAN_WEEKLY_WAGES ~ GDP + UNRATE, data=DF_indicators)</pre>
summary(model2)
```

lm(formula = MEDIAN_WEEKLY_WAGES ~ GDP + UNRATE, data = DF_indicators)

```
##
## Residuals:
##
       Min
                1Q Median
                                       Max
## -12.865 -3.693 -1.033
                             3.978
                                    17.201
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) -5.52624
                          49.89618
                                    -0.111
                                               0.912
## GDP
                0.04086
                           0.00204
                                    20.026 < 2e-16 ***
## UNRATE
               13.82381
                           2.24317
                                     6.163 5.98e-07 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 7.348 on 33 degrees of freedom
## Multiple R-squared: 0.9841, Adjusted R-squared: 0.9832
## F-statistic: 1024 on 2 and 33 DF, p-value: < 2.2e-16
model3 <- lm(MEDIAN_WEEKLY_WAGES ~ GDP + HOUST, data=DF_indicators)</pre>
summary(model3)
##
## Call:
## lm(formula = MEDIAN_WEEKLY_WAGES ~ GDP + HOUST, data = DF_indicators)
##
## Residuals:
##
       Min
                1Q Median
                                3Q
                                        Max
## -15.734 -6.178 -0.912
                             5.764 35.614
##
## Coefficients:
                 Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 254.963578 24.932722 10.226 9.22e-12 ***
## GDP
                 0.033336
                            0.002336 14.270 1.13e-15 ***
## HOUST
                -0.013550
                            0.006645
                                      -2.039
                                                0.0495 *
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 10.16 on 33 degrees of freedom
## Multiple R-squared: 0.9697, Adjusted R-squared: 0.9679
## F-statistic:
                  528 on 2 and 33 DF, p-value: < 2.2e-16
Each second predictor (DJCA, UNRATE, HOUST) has a p-value < 0.15, and is thus a candidate to enter the
model. We choose to proceed with the one with the lowest p-value, which is UNRATE. Note that our p-value
for GDP remains below 0.15, so we proceed with the MEDIAN_WEEKLY_WAGES \sim GDP + UNRATE
model.
  3. Fit each three-predictor model that includes GDP and UNRATE.
model1 <- lm(MEDIAN_WEEKLY_WAGES ~ GDP + UNRATE + DJCA, data=DF_indicators)</pre>
summary(model1)
##
## Call:
## lm(formula = MEDIAN_WEEKLY_WAGES ~ GDP + UNRATE + DJCA, data = DF_indicators)
##
## Residuals:
##
        Min
                  1Q
                       Median
                                     3Q
                                             Max
```

```
## -13.6477 -3.3376 -0.0371 2.9161 17.6972
##
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) -61.818877 61.692482 -1.002
                            0.004310 10.810 3.26e-12 ***
## GDP
                0.046589
## UNRATE
                                      6.084 8.53e-07 ***
               13.470870
                            2.214328
## DJCA
               -0.007391
                           0.004926 - 1.501
                                                0.143
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 7.212 on 32 degrees of freedom
## Multiple R-squared: 0.9852, Adjusted R-squared: 0.9838
                 709 on 3 and 32 DF, p-value: < 2.2e-16
## F-statistic:
model2 <- lm(MEDIAN_WEEKLY_WAGES ~ GDP + UNRATE + HOUST, data=DF_indicators)</pre>
summary(model2)
##
## Call:
## lm(formula = MEDIAN_WEEKLY_WAGES ~ GDP + UNRATE + HOUST, data = DF_indicators)
##
## Residuals:
##
       Min
                  1Q
                      Median
                                    3Q
                                            Max
## -12.8922 -4.0787 -0.2048
                                4.0326 17.8333
##
## Coefficients:
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) -20.155789 53.014829 -0.380
                                                0.706
## GDP
                 0.040389
                            0.002124 19.012 < 2e-16 ***
## UNRATE
                15.147181
                            2.743078
                                       5.522 4.35e-06 ***
## HOUST
                0.004972
                            0.005880
                                       0.846
                                                0.404
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 7.38 on 32 degrees of freedom
## Multiple R-squared: 0.9845, Adjusted R-squared: 0.983
## F-statistic: 676.7 on 3 and 32 DF, p-value: < 2.2e-16
Since the p-value for HOUST is greater than 0.15, it cannot be added to the model. We thus proceed with
MEDIAN_WEEKLY_WAGES \sim GDP + UNRATE + DJCA.
  4. Fit each four-predictor model that includes GDP, UNRATE, and DJCA.
model <- lm(MEDIAN_WEEKLY_WAGES ~ GDP + UNRATE + DJCA + HOUST, data=DF_indicators)</pre>
summary(model)
##
## Call:
## lm(formula = MEDIAN_WEEKLY_WAGES ~ GDP + UNRATE + DJCA + HOUST,
       data = DF_indicators)
##
##
## Residuals:
##
       Min
                  1Q
                      Median
                                    3Q
                                            Max
## -12.3470 -4.1344 -0.1523
                               3.5106 18.1108
##
```

```
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) -68.095184 63.190012 -1.078
## GDP
                0.045830
                           0.004533 10.110 2.47e-11 ***
## UNRATE
                14.439271
                           2.758068
                                      5.235 1.09e-05 ***
                -0.006848
                           0.005058
## DJCA
                                     -1.354
                                               0.186
                0.003541
## HOUST
                           0.005900
                                      0.600
                                               0.553
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 7.285 on 31 degrees of freedom
## Multiple R-squared: 0.9853, Adjusted R-squared: 0.9835
## F-statistic: 521.2 on 4 and 31 DF, p-value: < 2.2e-16
```

Note that the p-value for HOUST is greater than our alpha_to_enter, so we do not include it in the model. Thus, our final regression model is as follows:

```
model_final <- lm(MEDIAN_WEEKLY_WAGES ~ GDP + UNRATE + DJCA, data=DF_indicators)
summary(model_final)</pre>
```

```
##
## Call:
## lm(formula = MEDIAN_WEEKLY_WAGES ~ GDP + UNRATE + DJCA, data = DF_indicators)
##
## Residuals:
##
       Min
                  1Q
                      Median
                                    ЗQ
                                            Max
## -13.6477 -3.3376 -0.0371
                                2.9161 17.6972
##
## Coefficients:
                Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) -61.818877
                          61.692482 -1.002
                                                0.324
                            0.004310 10.810 3.26e-12 ***
## GDP
                0.046589
## UNRATE
                13.470870
                            2.214328
                                      6.084 8.53e-07 ***
## DJCA
                -0.007391
                            0.004926
                                     -1.501
                                                0.143
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 7.212 on 32 degrees of freedom
## Multiple R-squared: 0.9852, Adjusted R-squared: 0.9838
## F-statistic:
                 709 on 3 and 32 DF, p-value: < 2.2e-16
```

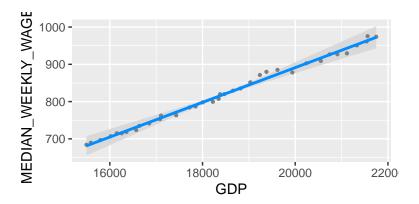
Regression Model and Interpretations

1. We have thus found our final regression model. In standard form:

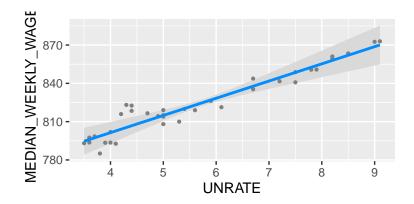
$$\widehat{\text{MEDIAN}}_{\text{WEEKLY}}_{\text{WAGES}} = -61.82 + 0.05(\text{GDP}) + 13.47(\text{UNRATE}) - 0.01(\text{DJCA})$$

- 2. Coefficient Interpretation.
- When UNRATE and DJCA are held constant, every one billion dollar increase in GDP increases median weekly wages by \$0.05.
- When GDP and DJCA are held constant, every one unit increase in unemployment increases median weekly wages by \$13.47.
- When GDP and UNRATE are held constant, every one unit increase in the DJCA index increases median weekly wages by \$0.01.

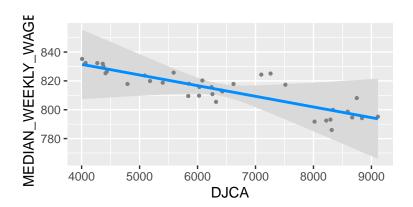
- 3. Conditional Relationship Plots
- GDP:



• UNRATE:



• DJCA:



4. R-squared.

```
##
## Call:
## lm(formula = MEDIAN_WEEKLY_WAGES ~ GDP + UNRATE + DJCA, data = DF_indicators)
##
## Residuals:
## Min 1Q Median 3Q Max
## -13.6477 -3.3376 -0.0371 2.9161 17.6972
##
```

```
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) -61.818877 61.692482 -1.002
## GDP
                           0.004310 10.810 3.26e-12 ***
                0.046589
## UNRATE
               13.470870
                           2.214328
                                      6.084 8.53e-07 ***
## DJCA
               -0.007391
                           0.004926
                                     -1.501
                                               0.143
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 7.212 on 32 degrees of freedom
## Multiple R-squared: 0.9852, Adjusted R-squared: 0.9838
                 709 on 3 and 32 DF, p-value: < 2.2e-16
## F-statistic:
```

As we see in the summary of the model, we have an R-squared of 0.9852. This means that our model explains 98.52% of the variance in MEDIAN WEEKLY WAGES.

Predictions and Future Research Questions

1. Given GDP = 21,494.731, UNRATE = 6.7, and DJCA = 9,732.69, we will use our model to predict MEDIAN WEEKLY WAGES.

```
NEWDATA = data.frame(GDP=21494.731, UNRATE=6.7, DJCA=9732.69)
predict(model_final, NEWDATA, interval="predict")

## fit lwr upr
## 1 957.9113 933.218 982.6046
```

As we can see in the R output, we compute a fitted value of \$957.91, with prediction interval of [\$933.22, \$982.60].

2. R-squared-adjusted.

```
summary(model_final)
```

```
##
## Call:
## lm(formula = MEDIAN_WEEKLY_WAGES ~ GDP + UNRATE + DJCA, data = DF_indicators)
##
## Residuals:
##
       Min
                  1Q
                      Median
                                    3Q
                                            Max
## -13.6477 -3.3376 -0.0371
                                2.9161
                                       17.6972
##
## Coefficients:
                 Estimate Std. Error t value Pr(>|t|)
##
                          61.692482 -1.002
## (Intercept) -61.818877
## GDP
                 0.046589
                            0.004310 10.810 3.26e-12 ***
## UNRATE
                13.470870
                            2.214328
                                       6.084 8.53e-07 ***
## DJCA
                -0.007391
                            0.004926
                                     -1.501
                                                0.143
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 7.212 on 32 degrees of freedom
## Multiple R-squared: 0.9852, Adjusted R-squared: 0.9838
                  709 on 3 and 32 DF, p-value: < 2.2e-16
## F-statistic:
```

As we can see in the summary for our model, we have an adjusted-R-squared of 0.9838. This does not have the same interpretation as R-squared; in other words, this does not say that our model accounts of 98.38%

of variance in MEDIAN_WEEKLY_WAGES. However, adjusted-R-squared can be useful in comparing different possible models with varying numbers of explanatory variables.