

Lab 3

1. Create a multivariate time series; perform any interpolations.

```
vars <- c("year", "conpress", "sex", "age", "degree", "wrkstat")
GSS <- data.table::fread("/Users/hengyuai/Desktop/TS/Lab3/trends-gss.csv",
  sep = ",",
  select = vars,
  data.table = FALSE)
```

```
sub <- GSS[, vars]
sub = as.data.frame(sub)
```

```
sub <- mutate(sub,
  trustpress = ifelse(conpress < 3, 1, 0),
  baplus = ifelse(degree >= 3, 1, 0),
  degreeelt50 = ifelse(baplus == 1 & age < 50, 1, 0),
  fulltime = ifelse(wrkstat == 1, 1, 0))
```

My QUESTION is: Are people's confidence in the press and their working status related over time in t

get means by year

```
by.year <- aggregate(subset(sub, sel = -year), list(year = sub$year), mean, na.rm = T)
```

interpolate for some missing years

add the extra years

```
by.year[30:40, "year"] <- c(1979, 1981, 1992, 1995, seq(1997, 2009, 2))
```

```
by.year <- arrange(by.year, year)
```

make a time series object by.year.ts and interpolate using na.approx

```
by.year.ts <- ts(by.year)
```

```
by.year.ts <- na.approx(by.year.ts)
```

calculate percent tatholic and percent under 50 with BA

```
by.year.ts <- as.data.frame(by.year.ts)
```

```
by.year.ts <- mutate(by.year.ts,
  fulltime_pct = fulltime*100,
  degreeelt50_pct = degreeelt50*100)
```

only keep up to 1992 and convert back to time series object

```
by.year.ts <- ts(subset(by.year.ts, year <= 1992))
```

correlations

```
cor.vars <- c("trustpress", "fulltime_pct", "degreeelt50_pct", "age", "year")
```

```
cor.dat <- by.year.ts[, cor.vars]
```

```
cor(cor.dat, use = "complete")
```

```
##               trustpress fulltime_pct degreeelt50_pct      age
## trustpress      1.0000000  -0.4804851  -0.9281833 -0.6562911
## fulltime_pct  -0.4804851   1.0000000   0.5796194  0.3590326
```

```
## degreeelt50_pct -0.9281833    0.5796194    1.0000000    0.6255061
## age            -0.6562911    0.3590326    0.6255061    1.0000000
## year          -0.9250063    0.6809984    0.9177727    0.7119450
##               year
## trustpress     -0.9250063
## fulltime_pct   0.6809984
## degreeelt50_pct 0.9177727
## age            0.7119450
## year           1.0000000
```

2. Graph the relationships between X and Y. Explain how you think Y should relate to your key Xs.

```
# Time series plots with ggplot
# install.packages("reshape2")

# Make a character vector naming the variables we might want to plot
keep.vars <- c("year", "trustpress", "fulltime_pct", "degreeelt50_pct", "age")

# Use meltMyTS to transform the data to a 3-column dataset containing a column for time, a column for v

library("reshape2")

meltMyTS <- function(mv.ts.object, time.var, keep.vars){
  # mv.ts.object = a multivariate ts object
  # keep.vars = character vector with names of variables to keep
  # time.var = character string naming the time variable
  require(reshape2)

  if(missing(keep.vars)) {
    melt.dat <- data.frame(mv.ts.object)
  }
  else {
    if (!(time.var %in% keep.vars)){
      keep.vars <- c(keep.vars, time.var)
    }
    melt.dat <- data.frame(mv.ts.object)[, keep.vars]
  }
  melt.dat <- melt(melt.dat, id.vars = time.var)
  colnames(melt.dat)[which(colnames(melt.dat) == time.var)] <- "time"
  return(melt.dat)
}

plot.dat <- meltMyTS(mv.ts.object = by.year.ts, time.var = "year", keep.vars = keep.vars)
plot.dat
```

```
##    time    variable    value
## 1  1972    trustpress      NA
## 2  1973    trustpress 0.8510494
## 3  1974    trustpress 0.8229665
## 4  1975    trustpress 0.8162275
## 5  1976    trustpress 0.8202324
```

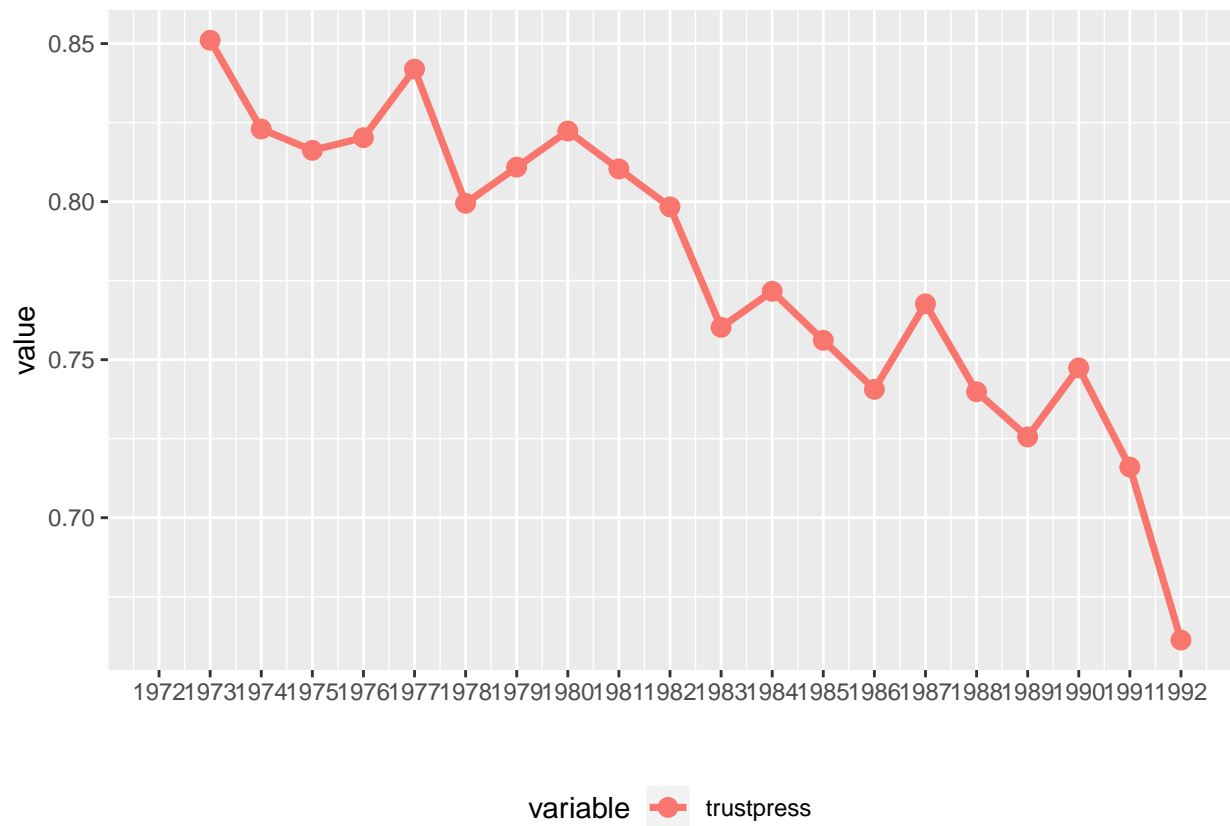
## 6	1977	trustpress	0.8419290
## 7	1978	trustpress	0.7994670
## 8	1979	trustpress	0.8108992
## 9	1980	trustpress	0.8223315
## 10	1981	trustpress	0.8103347
## 11	1982	trustpress	0.7983380
## 12	1983	trustpress	0.7602302
## 13	1984	trustpress	0.7716371
## 14	1985	trustpress	0.7561245
## 15	1986	trustpress	0.7406120
## 16	1987	trustpress	0.7676653
## 17	1988	trustpress	0.7398543
## 18	1989	trustpress	0.7255489
## 19	1990	trustpress	0.7474048
## 20	1991	trustpress	0.7160121
## 21	1992	trustpress	0.6613006
## 22	1972	fulltime_pct	46.4972102
## 23	1973	fulltime_pct	43.2845745
## 24	1974	fulltime_pct	42.7897574
## 25	1975	fulltime_pct	41.7449664
## 26	1976	fulltime_pct	41.2941961
## 27	1977	fulltime_pct	50.6535948
## 28	1978	fulltime_pct	46.7362924
## 29	1979	fulltime_pct	46.7673288
## 30	1980	fulltime_pct	46.7983651
## 31	1981	fulltime_pct	45.9529460
## 32	1982	fulltime_pct	45.1075269
## 33	1983	fulltime_pct	46.3414634
## 34	1984	fulltime_pct	48.6761711
## 35	1985	fulltime_pct	48.3050847
## 36	1986	fulltime_pct	47.4829932
## 37	1987	fulltime_pct	51.3468939
## 38	1988	fulltime_pct	49.1559757
## 39	1989	fulltime_pct	49.4469746
## 40	1990	fulltime_pct	51.3119534
## 41	1991	fulltime_pct	46.2755438
## 42	1992	fulltime_pct	48.3245714
## 43	1972	degreeelt50_pct	7.6971214
## 44	1973	degreeelt50_pct	9.7855228
## 45	1974	degreeelt50_pct	10.2564103
## 46	1975	degreeelt50_pct	9.8723976
## 47	1976	degreeelt50_pct	10.1808439
## 48	1977	degreeelt50_pct	9.7769029
## 49	1978	degreeelt50_pct	10.1894187
## 50	1979	degreeelt50_pct	10.5554602
## 51	1980	degreeelt50_pct	10.9215017
## 52	1981	degreeelt50_pct	10.2367897
## 53	1982	degreeelt50_pct	9.5520777
## 54	1983	degreeelt50_pct	12.2180451
## 55	1984	degreeelt50_pct	13.1793478
## 56	1985	degreeelt50_pct	12.3939987
## 57	1986	degreeelt50_pct	13.9740968
## 58	1987	degreeelt50_pct	14.0650855
## 59	1988	degreeelt50_pct	13.5043889

```
## 60 1989 degreeelt50_pct 13.9124755
## 61 1990 degreeelt50_pct 13.5865595
## 62 1991 degreeelt50_pct 15.1655629
## 63 1992 degreeelt50_pct 15.7963979
## 64 1972          age 44.9508706
## 65 1973          age 44.1820000
## 66 1974          age 44.5913396
## 67 1975          age 44.3077441
## 68 1976          age 45.2866711
## 69 1977          age 44.6631648
## 70 1978          age 44.0098361
## 71 1979          age 44.4922381
## 72 1980          age 44.9746402
## 73 1981          age 44.9168594
## 74 1982          age 44.8590786
## 75 1983          age 44.2964824
## 76 1984          age 44.0047716
## 77 1985          age 45.7111984
## 78 1986          age 45.4306220
## 79 1987          age 44.9236303
## 80 1988          age 45.3744076
## 81 1989          age 45.4435747
## 82 1990          age 45.9569971
## 83 1991          age 45.6261559
## 84 1992          age 45.8374377
```

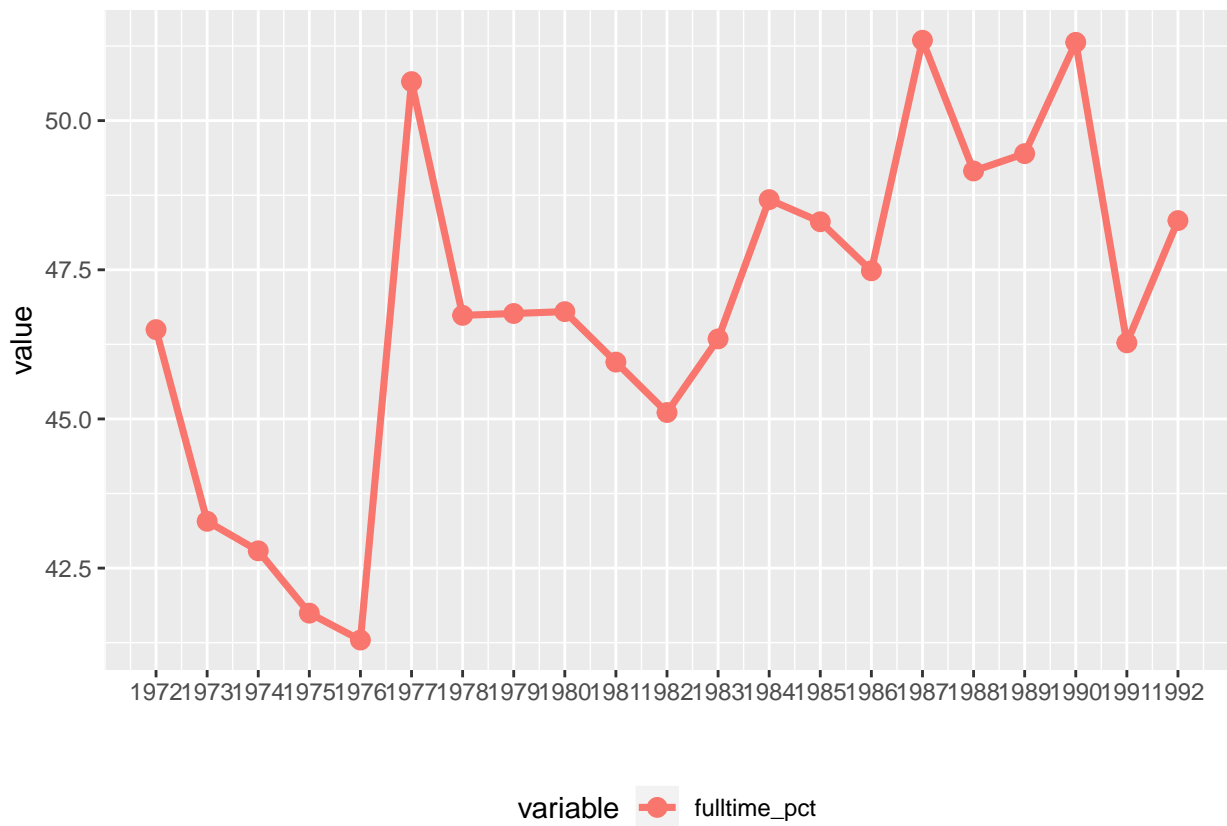
```
ggMyTS <- function(df, varlist, line = TRUE, point = TRUE, pointsize = 3, linewidth = 1.25, ...){
  require(ggplot2)
  # varlist = character vector with names of variables to use
  if(missing(varlist)){
    gg <- ggplot(df, aes(time, value, colour = variable))
  }
  else{
    include <- with(df, variable %in% varlist)
    gg <- ggplot(df[include,], aes(time, value, colour = variable))
  }
  if(line == FALSE & point == FALSE) {
    stop("At least one of 'line' or 'point' must be TRUE")
  }
  else{
    if(line == TRUE) gg <- gg + geom_line(size = linewidth, aes(color = variable), ...)
    if(point == TRUE) gg <- gg + geom_point(size = pointsize, aes(color = variable), ...)
  }

  gg + xlab("") + theme(legend.position = "bottom") + scale_x_continuous(breaks = min(df$time):max(df$time))
}

(g_trustpress <- ggMyTS(df = plot.dat, varlist = c("trustpress")))
```



```
(g_tvholipect <- ggMyTS(df = plot.dat, varlist = c("fulltime_pct")))
```



```
(g_degreeelt50_pct <- ggMyTS(df = plot.dat, varlist = c("degreeelt50_pct")))
```



Explain how you think Y should relate to your key Xs: From the graphs above, we can find that the percentage of people who have fulltime jobs and the percentage of people who with at least a BA increased between 1972 and 1992 overall. However, people's average confidence in press declined. Therefore, I think that people's average confidence in press was negatively related to the percentage of people with fulltime jobs and a BA under 50.

3. Run a simple time series regression, with one X and no trend. Interpret it.

```
# simplest regression
lm.trust <- lm(trustpress ~ fulltime_pct, data = by.year.ts)
summary(lm.trust)
```

```
##
## Call:
## lm(formula = trustpress ~ fulltime_pct, data = by.year.ts)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.10638 -0.02173  0.00428  0.02386  0.09262
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   1.149076   0.159507   7.204 1.05e-06 ***
## fulltime_pct -0.007892   0.003395  -2.324  0.032 *
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.04365 on 18 degrees of freedom
## (1 observation deleted due to missingness)
## Multiple R-squared:  0.2309, Adjusted R-squared:  0.1881
## F-statistic: 5.403 on 1 and 18 DF,  p-value: 0.032
```

The percent people working full-time was negatively related to average confidence in press. The coefficient is statistically significant at 0.05 and we can reject the null of no effect.

```
# test for heteroskedasticity
bptest(lm.trust)
```

```
##
## studentized Breusch-Pagan test
##
## data:  lm.trust
## BP = 1.3552, df = 1, p-value = 0.2444
```

There is no heteroskedasticity from the above regression.

4. Run a time series regression with one X and trend. Interpret it. Perform autocorrelation diagnostics. Explain what you found.

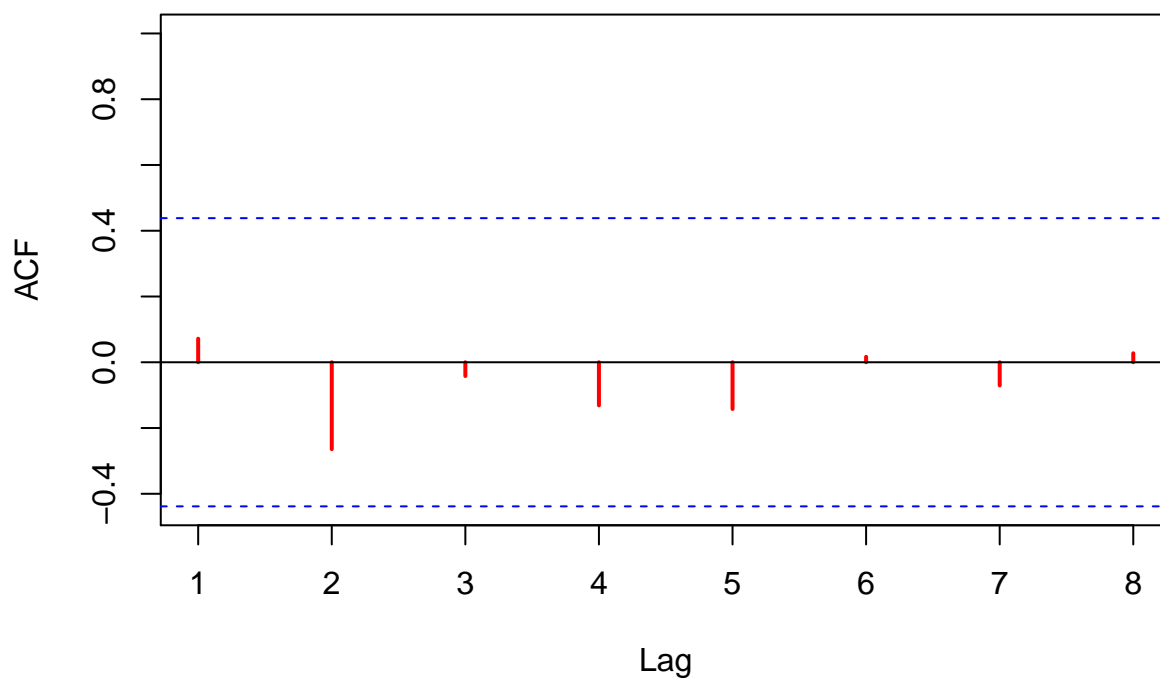
```
lm.trust2 <- update(lm.trust, ~ . + year)
summary(lm.trust2)
```

```
##
## Call:
## lm(formula = trustpress ~ fulltime_pct + year, data = by.year.ts)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.037563 -0.008056 -0.000676  0.011149  0.022925
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  18.6597301   1.6684148   11.184 2.93e-09 ***
## fulltime_pct    0.0045776   0.0017436    2.625  0.0177 *
## year          -0.0091275   0.0008691  -10.502 7.51e-09 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.01641 on 17 degrees of freedom
## (1 observation deleted due to missingness)
## Multiple R-squared:  0.8973, Adjusted R-squared:  0.8852
## F-statistic: 74.25 on 2 and 17 DF,  p-value: 3.971e-09
```

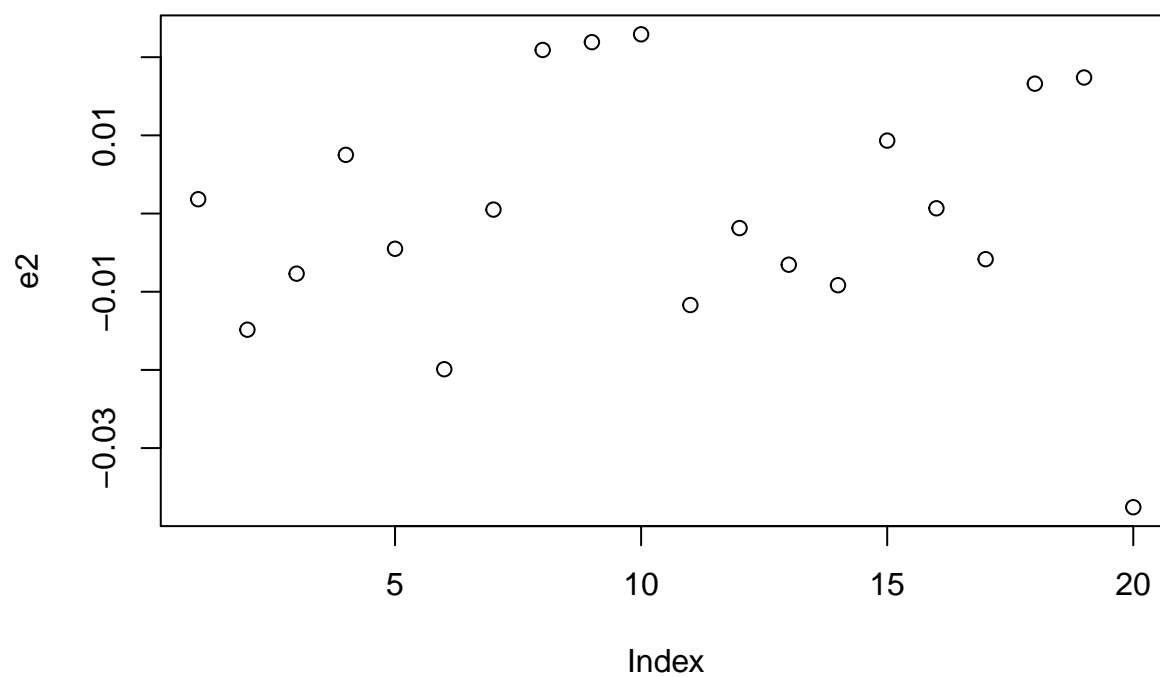
Net of the year trend, each percent more of full-time employed people increases ave. confidence in press by 0.0045776. This coefficient is significant at 0.05 level and we can reject the null of no effects.


```
# look for autocorrelation  
e2 <- lm.trust2$resid  
acf(e2, xlim = c(1,8), col = "red", lwd = 2)
```

Series e2



```
plot(e2)
```



```
dwtest(lm.trust2)
```

```
##
## Durbin-Watson test
##
## data:  lm.trust2
## DW = 1.5479, p-value = 0.08736
## alternative hypothesis: true autocorrelation is greater than 0
```

```
bgtest(lm.trust2)
```

```
##
## Breusch-Godfrey test for serial correlation of order up to 1
##
## data:  lm.trust2
## LM test = 0.17355, df = 1, p-value = 0.677
```

```
durbinWatsonTest(lm.trust2, max.lag=3)
```

```
## lag Autocorrelation D-W Statistic p-value
## 1 0.07160435 1.547916 0.192
## 2 -0.26429238 2.105385 0.842
## 3 -0.04201482 1.587673 0.492
## Alternative hypothesis: rho[lag] != 0
```

From the ACF graph and residual trend graph, we cannot see any AR(1) left. In the dwtest and bgtest result, a prob of $\chi^2 > 0.05$ indicates no serial correlation.

5. Consider running a time series regression with many Xs and trend. Interpret that. Check VIF.

```
lm.trust3 <- update(lm.trust2, ~ . + degreeelt50_pct)
summary(lm.trust3)
```

```
##
## Call:
## lm(formula = trustpress ~ fulltime_pct + year + degreeelt50_pct,
## data = by.year.ts)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.030039 -0.006378 -0.001098  0.009174  0.020660
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  11.958147   3.053569   3.916  0.00123 **
## fulltime_pct    0.003976   0.001543   2.576  0.02029 *
## year          -0.005671   0.001579  -3.592  0.00244 **
## degreeelt50_pct -0.010286   0.004117  -2.498  0.02376 *
```

```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.01435 on 16 degrees of freedom
## (1 observation deleted due to missingness)
## Multiple R-squared:  0.9261, Adjusted R-squared:  0.9123
## F-statistic: 66.84 on 3 and 16 DF,  p-value: 2.869e-09
```

Net of the year trend, each percent more of full-time employed people increases ave. confidence in press by 0.003976. This coefficient is significant at 0.05 level and we can reject the null of no effects. Net of the year trend, each percent more of under 50 BA degree people decreases ave. confidence in press by 0.010286. This coefficient is significant at 0.05 level and we can reject the null of no effects.

```
vif(lm.trust3) # variance inflation factor
```

```
##      fulltime_pct      year degreeelt50_pct
##      1.911386      8.048782      6.499728
```

Given such high correlations among variables, we want to look out for multicollinearity, which we might have with year and % of people under 50 with a BA+ degree.

6. Run a first differenced time series regression. Interpret that.

```
firstD <- function(var, group, df){
  bad <- (missing(group) & !missing(df))
  if (bad) stop("if df is specified then group must also be specified")

  fD <- function(j){ c(NA, diff(j)) }

  var.is.alone <- missing(group) & missing(df)

  if (var.is.alone) {
    return(fD(var))
  }
  if (missing(df)){
    V <- var
    G <- group
  }
  else{
    V <- df[, deparse(substitute(var))]
    G <- df[, deparse(substitute(group))]
  }

  G <- list(G)
  D.var <- by(V, G, fD)
  unlist(D.var)
}

by.yearFD <- summarise(data.frame(by.year.ts),
  trustpress = firstD(trustpress), # using firstD function from QMSS package
  age = firstD(age),
```

```

        fulltime_pct = firstD(fulltime_pct),
        degreeelt50_pct = firstD(degreeelt50_pct),
        year = year)

lm.trust4 <- update(lm.trust3, data = by.yearFD)
summary(lm.trust4)

##
## Call:
## lm(formula = trustpress ~ fulltime_pct + year + degreeelt50_pct,
##     data = by.yearFD)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.045549 -0.011698 -0.001073  0.015786  0.024447
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   0.8948219   1.7097579   0.523  0.6084
## fulltime_pct   0.0039378   0.0015902   2.476  0.0257 *
## year          -0.0004557   0.0008623  -0.528  0.6049
## degreeelt50_pct -0.0066733   0.0054022  -1.235  0.2357
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.02022 on 15 degrees of freedom
## (2 observations deleted due to missingness)
## Multiple R-squared:  0.3871, Adjusted R-squared:  0.2646
## F-statistic: 3.159 on 3 and 15 DF,  p-value: 0.05572

```

For each 1 percentage point change in people working-full time, average confidence in press increases by 0.0039378, net of all other differences in the Xs and at any point in time. This coefficient is significant at 0.05 level and we can reject the null of no effects. For each 1 percentage point change in people getting BA, average confidence in press decreases by 0.0066733, net of all other differences in the Xs and at any point in time. This coefficient is not significant.

7. Check your variables for unit roots. Do some tests. Interpret them.

```

# install.packages("fUnitRoots")
library(fUnitRoots)

## Loading required package: timeDate

## Loading required package: timeSeries

##
## Attaching package: 'timeSeries'

## The following object is masked from 'package:zoo':
##
##     time<-

```

```
## Loading required package: fBasics

##
## Attaching package: 'fBasics'

## The following object is masked from 'package:car':
##
##      densityPlot

adfTest(by.year.ts[, "trustpress"], lags = 0, type="ct")
```

```
##
## Title:
##   Augmented Dickey-Fuller Test
##
## Test Results:
##   PARAMETER:
##     Lag Order: 0
##   STATISTIC:
##     Dickey-Fuller: -2.842
##   P VALUE:
##     0.2516
##
## Description:
##   Fri Nov 29 01:27:25 2019 by user:
```

```
adfTest(by.year.ts[, "trustpress"], lags = 4, type="ct")
```

```
##
## Title:
##   Augmented Dickey-Fuller Test
##
## Test Results:
##   PARAMETER:
##     Lag Order: 4
##   STATISTIC:
##     Dickey-Fuller: -1.6725
##   P VALUE:
##     0.6971
##
## Description:
##   Fri Nov 29 01:27:25 2019 by user:
```

Either with 0 lag or with 4 lags, p-value is too high to be able to reject the null of Unit Root, therefore, we might have a unit root here.

8. Perform an Automatic ARIMA on the residuals from one of your earlier models. Tell me what it says.

```
library(forecast)
```

```
## Registered S3 method overwritten by 'xts':  
##   method      from  
##   as.zoo.xts zoo  
  
## Registered S3 method overwritten by 'quantmod':  
##   method      from  
##   as.zoo.data.frame zoo  
  
## Registered S3 methods overwritten by 'forecast':  
##   method      from  
##   fitted.fracdiff fracdiff  
##   residuals.fracdiff fracdiff
```

```
e <- lm.trust$resid  
auto.arima(e, trace=TRUE)
```

```
##  
## ARIMA(2,1,2) with drift      : Inf  
## ARIMA(0,1,0) with drift     : -63.006  
## ARIMA(1,1,0) with drift     : -63.08589  
## ARIMA(0,1,1) with drift     : Inf  
## ARIMA(0,1,0)                : -64.81855  
## ARIMA(1,1,1) with drift     : Inf  
##  
## Best model: ARIMA(0,1,0)  
  
## Series: e  
## ARIMA(0,1,0)  
##  
## sigma^2 estimated as 0.001717: log likelihood=33.53  
## AIC=-65.05 AICc=-64.82 BIC=-64.11
```

auto.arima suggests that the errors from the static model is a random walk and we cannot reject unit root.

9. Run an ARIMA that follows from Step 7. Interpret that, too.

```
xvars.fat <- by.year.ts[,c("fulltime_pct")]  
  
arima.010 <- arima(by.year.ts[, "trustpress"], order = c(0,1,0), xreg = xvars.fat)  
summary(arima.010)
```

```
##  
## Call:  
## arima(x = by.year.ts[, "trustpress"], order = c(0, 1, 0), xreg = xvars.fat)  
##  
## Coefficients:
```

```

##          xvars.fat
##          0.0039
## s.e.      0.0017
##
## sigma^2 estimated as 0.0004906:  log likelihood = 45.43,  aic = -86.86
##
## Training set error measures:
##              ME          RMSE          MAE          MPE          MAPE
## Training set -0.01044549 0.02158966 0.01621065 -1.425352 2.157985
##              MASE          ACF1
## Training set 0.7558092 -0.1133399

```

Each 1 percentage point difference in the percent of people with full-time job increases people's confidence in press by 0.0039 percentage points.