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* Analysis Script
* Murphy John
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* This script performs the main analyses of this project.
title "Setup";
** footnote;
footnote "Analysis script run on &SYSDATE at &SYSTIME.";
** establish library;
libname mylib "/home/u63984496/BIOS7400/final-project";
** set graphics ods;
ods graphics on / width=8in height=4in;
                           *****************
title "Exploratory analysis";
* plot the outcome of interest, avgprice;
** sort data by type, date, and region;
proc sort data=mylib.dat;
   by type date region;
** plot price over time by type stratified by region;
proc sgpanel data=mylib.dat;
   panelby type;
   series x = date y = avgprice / group = region;
run;
** plot price by pres_party and type stratified by region;
proc sgpanel data=mylib.dat;
   panelby region type;
   hbox avgprice / group = pres_party;
run:
** plot price by temp and type stratified by region;
proc sgpanel data=mylib.dat;
   panelby type;
   scatter x = temp y = avgprice / group = region;
* univariable analysis;
** get means for each type;
proc univariate data=mylib.dat plots;
   var avgprice;
   class type;
run;
* bivariable analysis;
** check correlations of numerical variables;
proc corr pearson data=mylib.dat;
   var temp avgprice;
   by type;
run;
** create macro to check means of categorical variables;
%macro means(var);
proc means data = mylib.dat;
   var avgprice;
   class &var;
   by type;
run;
%mend means:
* run;
%means(month num);
%means(year);
%means(region);
%means(pres_party);
```

```
title "Regression Model fits";
* create train and test data;
** randomly select 80 percent of the data for the training and reserve the remainder for testing;
proc surveyselect data=mylib.dat
    out=dat_select
    samprate=0.8
    outal1
    seed=333;
run;
** create seperate data sets;
data dat_train dat_test;
    set dat_select;
    if selected then output dat_train;
    else output dat_test;
run:
* simple linear regression;
** avgprice by type, only;
proc glm data=dat_train plots=all;
    class type (ref='conventional');
    model avgprice = type;
run:
*** rmse = 0.251153, rsq = 0.485679;
* bivariable regression with interaction;
** model avgprice by type, covariate, and interaction;
** covariates are month_num, year, temp, region, pres_party;
* write macro to fit the numerical covariate model;
%macro bivariable_num(covariate);
    proc glm data=dat_train plots=all;
        class type (ref='conventional');
        model avgprice = type &covariate type*&covariate;
    run;
%mend bivariable_num;
* write macro to fit the categorical covariate models;
%macro bivariable cat(covariate);
    proc glm data=dat_train plots=all;
        class type (ref='conventional') &covariate;
        model avgprice = type &covariate type*&covariate;
    run:
%mend bivariable_cat;
* run;
%bivariable_num(temp);
*** rmse = 0.234790, rsq = 0.552579;
%bivariable_cat(region);
*** rmse = 0.202991, rsq = 0.673291;
%bivariable_cat(pres_party);
*** rmse = 0.244407, rsq = 0.515176;
%bivariable_cat(year);
*** rmse = 0.239643, rsq = 0.538200;
%bivariable_cat(month_num);
*** rmse = 0.231129, rsq = 0.586450;
* create table of metrics by hand;
data stats;
    length Covariate $30;
    infile datalines dsd truncover;
    input Covariate :$30. R_Square RMSE;
    datalines;
"Type only",0.485679,0.251153
"Temperature",0.552579,0.234790
"Region", 0.673291, 0.202991
"Presidential Party", 0.515176, 0.244407
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"Year",0.538200,0.239643
"Month", 0.586450, 0.231129
run;
proc print data=stats noobs;
   var Covariate RMSE R Square;
*** All three bivariable models improve the fit above the univariable model as measued by the rmse and r-squared.
*** The bivariable model with region has the lowest rmse and highest r-squared.;
* full model;
** avgprice by type, temp, region, pres_party and interactions;
proc glm data=dat_train plots=all;
        class pres_party type (ref='conventional') region month_num year;
        model avgprice = type temp region pres_party month_num year
              type*temp type*region type*pres_party type*month_num type*year;
*** rmse = 0.145137, rsq = 0.844828;
* reduced model;
** the interaction between type and temp, month_num, and pres_party have small Type III SS and large p-values.
** Reduce the model by removing these terms.;
proc glm data=dat_train plots=all alpha=0.05;
        class type (ref='conventional') region pres_party month_num year;
        model avgprice = type temp region pres_party month_num year
              type*region;
        store out=final_model;
run:
quit;
*** rmse = 0.152099, rsq = 0.823078;
*** The reduced model is very similar in terms of fit to the full model but does not include unnessecary interaction terms.
*** Select this as the final model.;
                                   *************
title 'Model Evaluation';
* use the final model to generate predictions on the test data;
proc plm restore=final model;
   score data=dat_test out=predictions predicted;
run:
* compute rmse and r-squared;
** get residuals;
data eval;
   set predictions;
   resid = avgprice - predicted;
    sq_resid = resid**2;
run:
* compute mean;
proc means data=eval noprint;
   var avgprice predicted sq_resid;
   output out=metrics
        mean(avgprice)=mean_y
        sum(sq_resid)=ss_res
        n=samples;
run;
* compute metrics;
data results;
   set metrics;
    ss_total = 0;
    do i = 1 to samples;
       set eval point=i nobs=n;
       ss_total + (avgprice - mean_y)**2;
    rmse = sqrt(ss_res / samples);
   rsq = 1 - (ss_res / ss_total);
    keep rmse rsq;
run:
proc print data=results;
   title "RMSE and R-squared on Test Data";
```

```
* residuals vs fitted plot;
proc sgplot data=eval;
    scatter x=predicted y=resid / markerattrs=(symbol=circlefilled color=black);
    refline 0 / axis=y lineattrs=(color=red pattern=shortdash);
    xaxis label="Fitted Values";
    yaxis label="Residuals";
    title "Residuals vs. Fitted Values";

run;

* predicted vs observed plot;
proc sgplot data=predictions;
    scatter x=predicted y=avgprice / markerattrs=(symbol=circlefilled color=black);
    lineparm x=0 y=0 slope=1 / lineattrs=(color=red pattern=shortdash);
    xaxis label="Predicted Values";
    yaxis label="Observed Values";
    title "Predicted vs. Observed Values";
run;

* END OF SCRIPT:
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