Brain.sas: Explanation of code

Goals of code:

- Producing a scatterplot matrix
- Fitting a multiple regression using proc glm
- Fitting a multiple regression using proc reg
- Regression diagnostics using proc reg

**Scatterplot matrix:** proc sgscatter; matrix;
We have used proc sgplot before to draw bivariate plots (Y vs X). sgplot can draw many different types of plot and other sg (statistical graphics) procedures draw more. proc sgscatter; matrix;
draws a scatterplot matrix of the variables specified after the keyword matrix and before the terminating ;

You can add any of the usual modifying commands (by, where) or annotation commands (title) to further control what gets plotted.

**multiple regression using proc glm** proc glm; model;
proc glm allows you to specify multiple X variables on the right hand side of the model statement. This will fit a multiple regression.

Lecture will discuss sequential (Type I) and partial (Type III) tests.

The numeric output includes:

- The ANOVA table comparing the full model (all X variables) to the intercept only model.
- Four numbers describing the full model: The most useful (for me) is Root MSE, which is the estimate of the residual standard deviation, i.e., the square root of the MSE from the ANOVA table.
- Type I (sequential) sums-of-squares and associated tests, for adding each term in the model
- Type III (partial) sums-of-squares and associated tests, for adding each term last to the model.
- Parameter estimates, their standard errors, and tests of each parameter = 0.

This may be followed by a plot of the model predictions. What sort of plot depends on the model being fit. This is a surface plot if the model has 2 X variables. No plot is produced when there are 3 or more X variables.
**multiple regression using proc reg**  

```
proc reg; model ;
```

`proc glm` can fit models with continuous X variables or variables that indicate groups (i.e. in a class statement). If you don’t have any grouping variables (no class statement needed), you can fit the model using `proc reg`. The results are identical, but the output is organized differently.

When we look at additional regression output, e.g. diagnostics, `proc reg` and `proc glm` provide overlapping but not identical options.

The `proc reg` syntax is identical to the `proc glm` syntax. Variable names for the model to be fit go on the right hand side of the equals sign in the model statement.

The numeric output includes:

- The overall ANOVA table, identical to that from `proc glm`
- A five number summary of the full model. Four numbers are the same as those in the `proc glm` output. Again, Root MSE is the number I find most useful.
- Parameter estimates, their standard errors, and tests of parameter = 0.

This is followed by a page of graphical diagnostics. The first one is the residual vs predicted values plot. We will talk about the most of the other diagnostics in the next week or so.

**Regression diagnostics:**   \r vif and output

If you want to see the actual numbers for regression diagnostics, adding the `/r` option to the `proc reg` prints out diagnostics for each observation.

The VIF (variance inflation factor) diagnostic is characteristic of each variable (not each observation). To print that out, add `/vif` to the `proc reg` statement.

Note: If you have two options, e.g. `/r` and `/vif`, you only need one `/` to separate the model piece from the options. This is illustrated in the example code in the last `proc reg`.

If you want to store the diagnostics in a data set, which can then be printed or plotted, use `output` with the appropriate keyword. The example code gives keywords for the diagnostics we’ll use in class and a few more that I’ll briefly mention.

Many diagnostics, e.g., Cook’s Distance, are best plotted against the observation number. This gives you a quick indication of the largest value (i.e., the answer to the question, do I need to care about influential observations?) and help you identify which point is the one (or more) with large D values.

**Storing the observation number in a data set:**   \i = _n_;

The easiest way to produce a full page plot using the observation number is to create a new variable
(I use \(i\)) containing the observation number. This is done in the data step that creates \texttt{brain2}. Inside a data step, SAS creates various “internal” variables. These have values that are defined inside the data step but are not saved, so they are not accessible to procs. When SAS creates a variable, it usually has \_ before and after, so the created variable is not confused with a variable name you want. One of those “internal” variables is \_\texttt{N}_\_. This contains the observation number. To save that information, copy it to a variable you name (e.g. \texttt{i}), which will then be stored and can used in subsequent procs.