

SAS

Most popular Statistical software worldwide.

SAS claims that its products are used at over 40,000 sites, including at 90% of the Fortune 500.

This will not be all SAS as they make other products, such as JMP (a menu and dialogue based stat package)

Huge in the pharmaceutical industry. There is a belief, though 100% FALSE, that the FDA requires the analysis of all clinical trials must be done if SAS.

You can do it in anything, just need to document it as part of your approval submission.

SAS, the company, has the reputation of being a fantastic place to work as well.

Extremely powerful package. You name it, it probably does it. If it doesn't, they are probably working on it.

It is available for many platforms including Windows, Unix, Macintosh, mainframes (z/OS, CMS, VSE, VMS, MVS)

However they don't keep all versions updated at the same rate.

It's a program based package. You need to write a program for your analysis.

These programs will have a block structure, with each block corresponding to a different part of the analysis.

Each block will usually start with a different PROC statement, such as PROC REGRESS, PROC SORT, PROC LOGISTIC, etc.

Within each block, command will be given, options set, etc.

Programming features

Case doesn't matter – PROC, proc, & PrOc all mean the same thing.

All commands must end with a ';'. This allows commands to be split across lines.

When commenting your code, you can either do it with

```
* comments ;
```

or

```
/* comments */
```

The following is a program which performs a one way Analysis of variance, does some calculations with the residuals, and then gives some diagnostic plots.

```
* Sample SAS program
* Data set is from Dean and Voss (1999) Design and Analysis of
* Experiments. Problem 3, page 129.

options linesize=75; /* set the output width to 75 characters */

data temp;
  infile 'p147.3';
  input brand time;
  invtime=1/time;

* print the data to see if everything is ok
proc print data=temp;
  title 'Margarine Experiment';

* Run the ANOVA
proc glm;
  classes brand;
  model invtime = brand;
  estimate 'marg vs but' brand 1 1 1 -3/divisor=3;
  output out=resids predicted=pred residual=z;
run;

* Switch from data file temp to data file resids
data;
  set resids;

*Standardize the residuals to have standard deviation 1
proc standard std=1.0;
  var z;

* Calculate the normal scores with Blom's adjustment
proc rank normal=blom;
  var z;
  ranks nscores;

proc print;

* Generate diagnostic plots
proc plot;
  plot invtime*brand z*pred z*brand z*nscores;

run;

quit; /* Ends the program */
```

This program will generate a lot of output. SAS tends to do that.

Also SAS will generate two files when running a program, your results (.lst) and a log file (.log), describing what commands have been run along with comments and error messages.

Also, the default file extension for programs is .sas.

However you can use any file extension you want.

With the results, the default is to return plain text.

It is possible to create your output in other formats, such as HTML (with the ODS command)

Margarine Experiment 41
17:01 Wednesday, December 10, 2003

Obs	brand	time	invtime
1	1	167	.005988024
2	1	171	.005847953
3	1	178	.005617978
4	1	175	.005714286
5	1	184	.005434783
6	1	176	.005681818
7	1	185	.005405405
8	1	172	.005813953
9	1	178	.005617978
10	1	178	.005617978
11	2	231	.004329004
12	2	233	.004291845
13	2	236	.004237288
14	2	252	.003968254
15	2	233	.004291845
16	2	225	.004444444
17	2	241	.004149378
18	2	248	.004032258
19	2	239	.004184100
20	2	248	.004032258
21	3	176	.005681818
22	3	168	.005952381
23	3	171	.005847953
24	3	172	.005813953
25	3	178	.005617978
26	3	176	.005681818
27	3	169	.005917160
28	3	164	.006097561
29	3	169	.005917160
30	3	171	.005847953
31	4	201	.004975124
32	4	199	.005025126
33	4	196	.005102041
34	4	211	.004739336

Margarine Experiment 42
17:01 Wednesday, December 10, 2003

Obs	brand	time	invtime
35	4	209	.004784689
36	4	223	.004484305
37	4	209	.004784689
38	4	219	.004566210
39	4	212	.004716981
40	4	210	.004761905

Margarine Experiment 43
17:01 Wednesday, December 10, 2003

The GLM Procedure

Class Level Information

Class	Levels	Values
brand	4	1 2 3 4

Number of observations 40

The GLM Procedure

Dependent Variable: invtime

Source	DF	Sum of Squares	Mean Square	F Value	Pr > F
Model	3	0.00001782	0.00000594	208.33	<.0001
Error	36	0.00000103	0.00000003		
Corrected Total	39	0.00001884			

R-Square	Coeff Var	Root MSE	invtime Mean
0.945537	3.294115	0.000169	0.005125

Source	DF	Type I SS	Mean Square	F Value	Pr > F
brand	3	0.00001782	0.00000594	208.33	<.0001

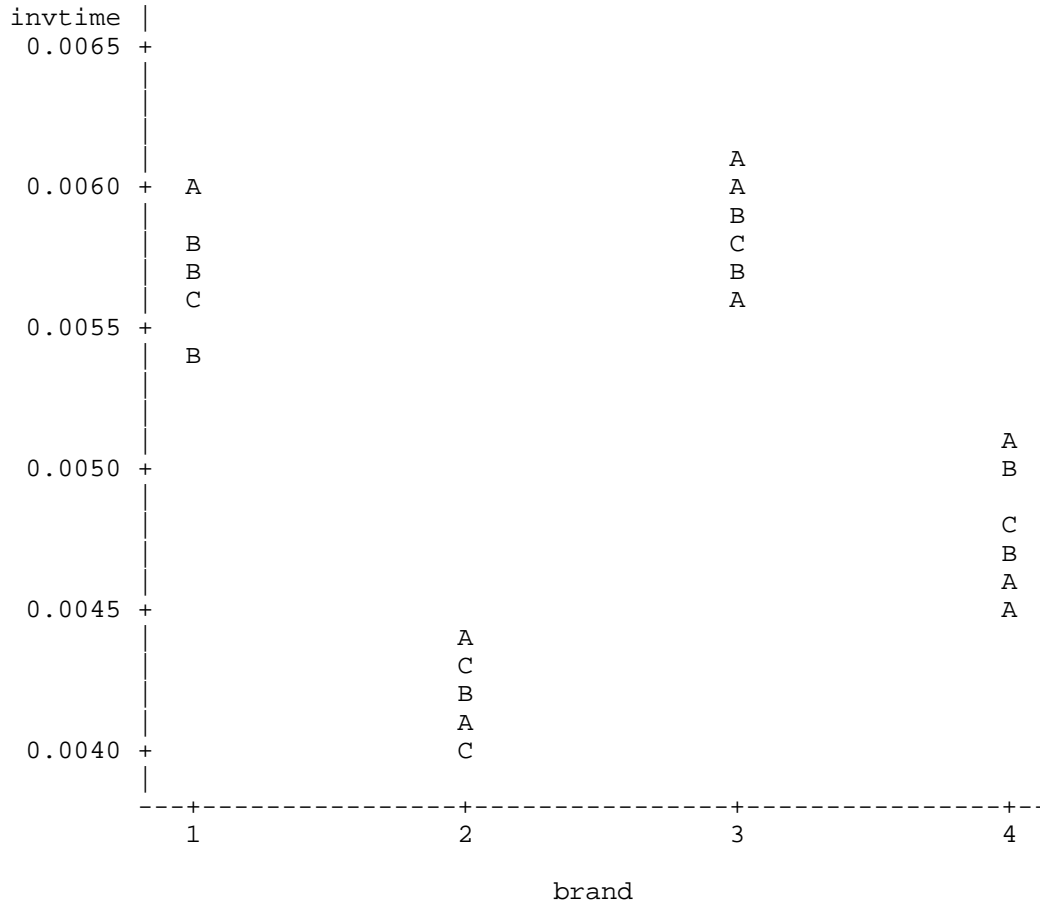
Source	DF	Type III SS	Mean Square	F Value	Pr > F
brand	3	0.00001782	0.00000594	208.33	<.0001

Parameter	Estimate	Standard Error	t Value	Pr > t
marg vs but	0.00044184	0.00006165	7.17	<.0001

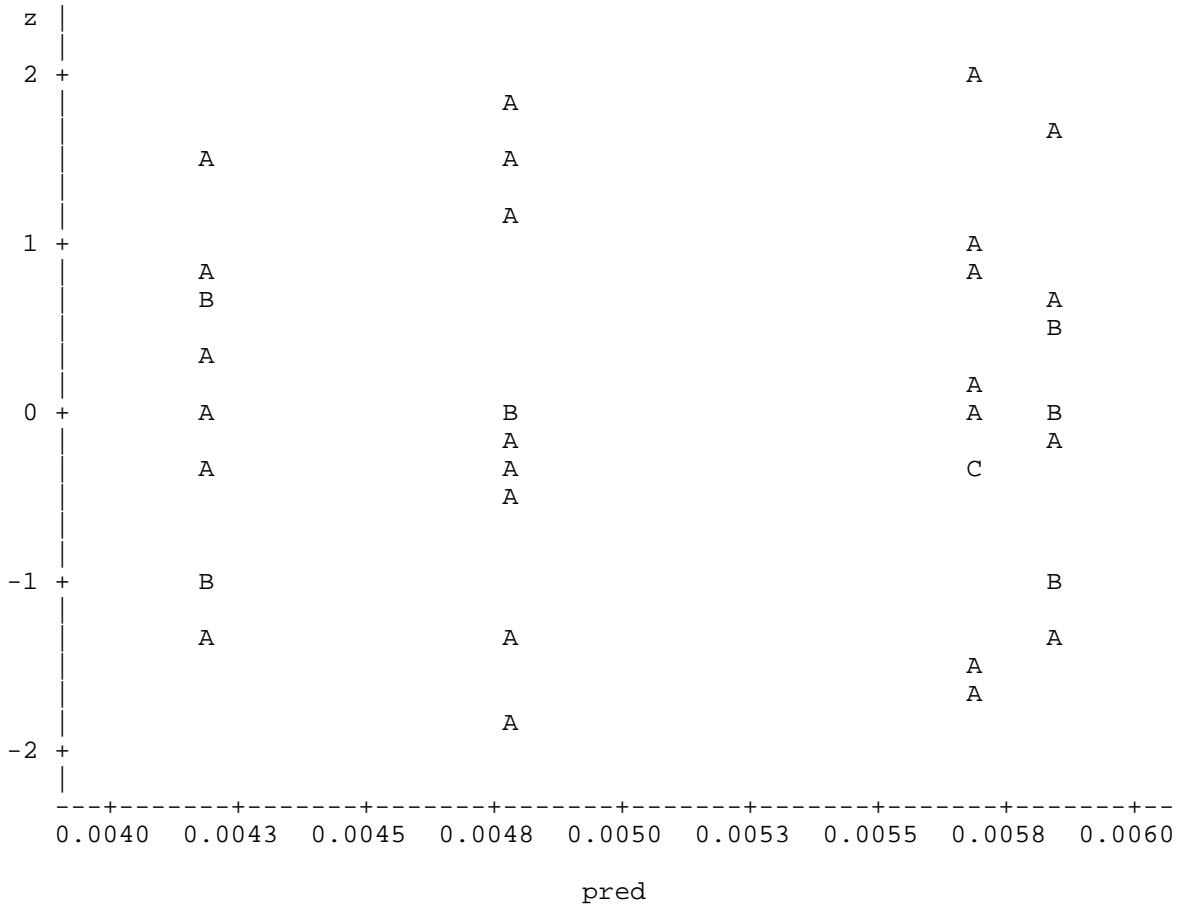
Obs	brand	time	invtime	pred	z	nscores
1	1	167	.005988024	.005674016	1.93577	2.15636
2	1	171	.005847953	.005674016	1.07227	0.97574
3	1	178	.005617978	.005674016	-0.34546	-0.48629
4	1	175	.005714286	.005674016	0.24825	0.28402
30	3	171	.005847953	.005837574	0.06399	0.18802
31	4	201	.004975124	.004794041	1.11633	1.08144
32	4	199	.005025126	.004794041	1.42457	1.20084
33	4	196	.005102041	.004794041	1.89873	1.74638
34	4	211	.004739336	.004794041	-0.33723	-0.34951

Obs	brand	time	invtime	pred	z	nscores
35	4	209	.004784689	.004794041	-0.05765	0.00000
36	4	223	.004484305	.004794041	-1.90943	-2.15636
37	4	209	.004784689	.004794041	-0.05765	0.00000
38	4	219	.004566210	.004794041	-1.40451	-1.34037
39	4	212	.004716981	.004794041	-0.47505	-0.63114
40	4	210	.004761905	.004794041	-0.19811	-0.21972

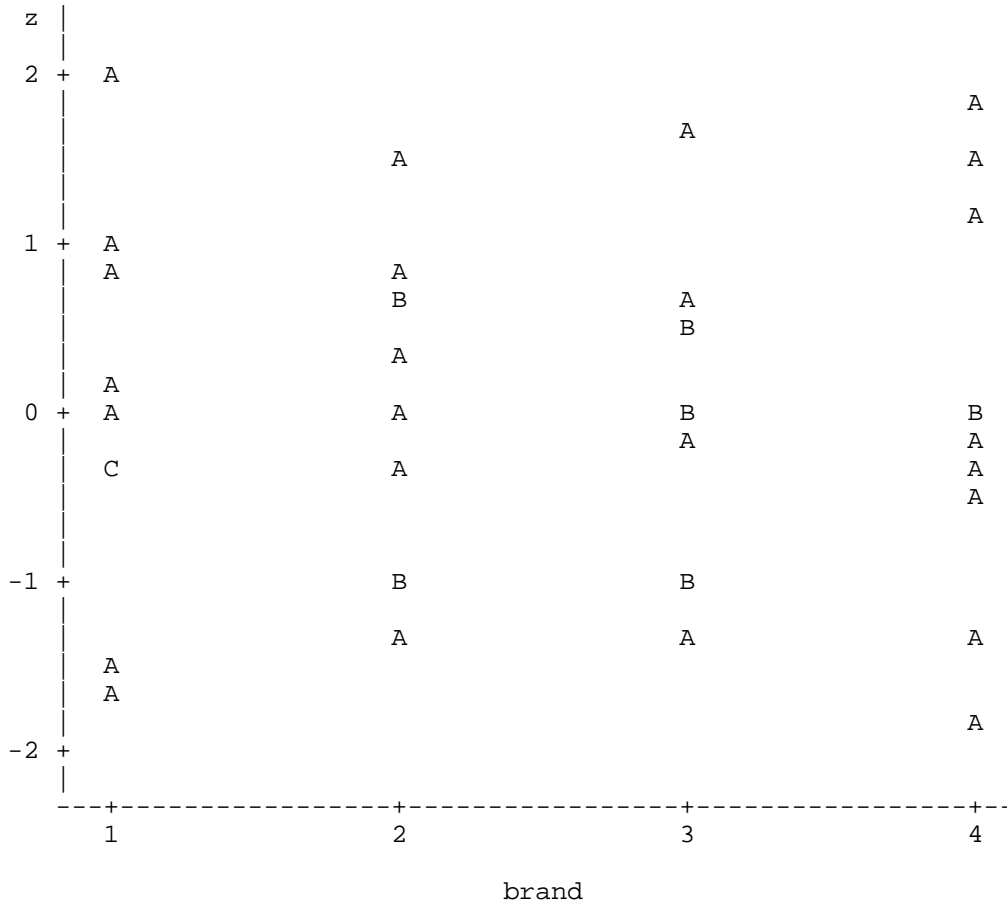
Plot of invtime*brand. Legend: A = 1 obs, B = 2 obs, etc.



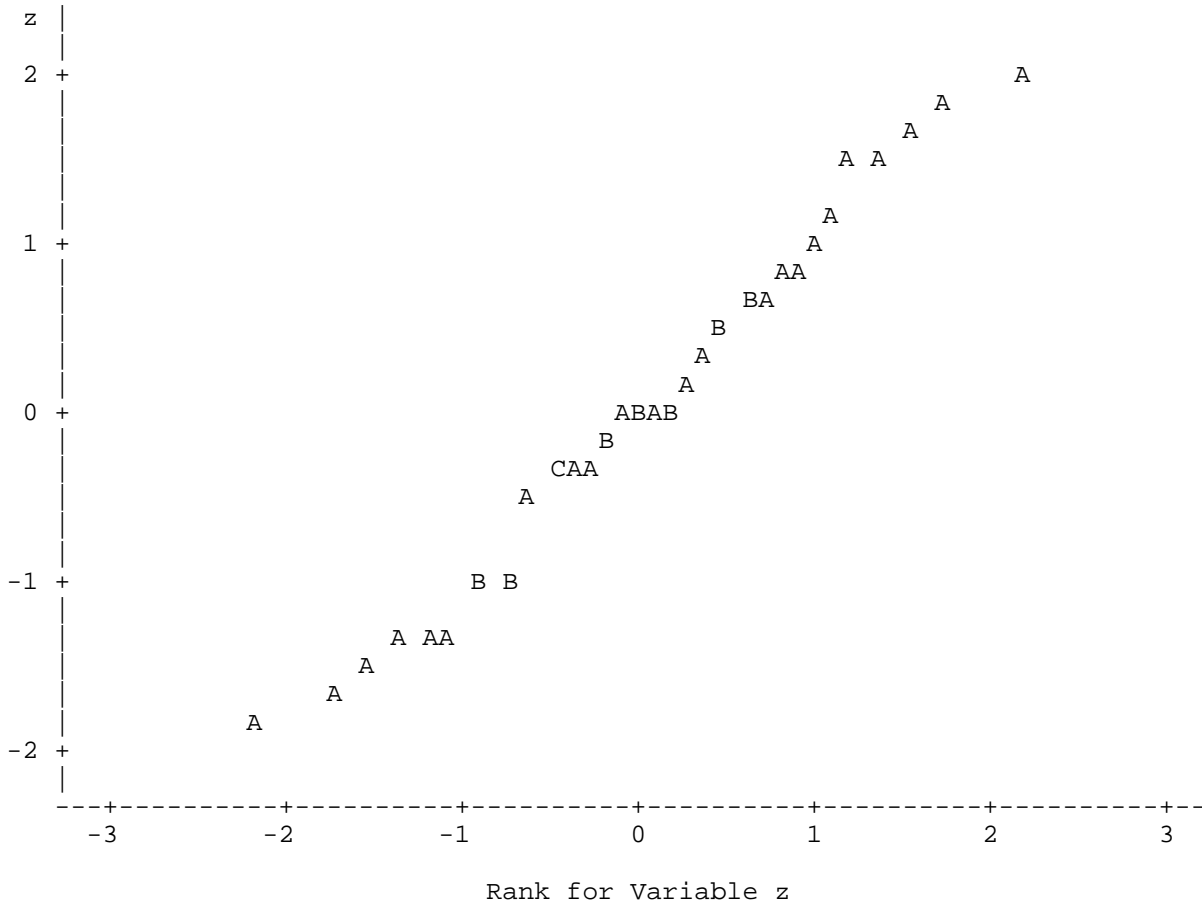
Plot of z*pred. Legend: A = 1 obs, B = 2 obs, etc.



Plot of z*brand. Legend: A = 1 obs, B = 2 obs, etc.



Plot of z*nscores. Legend: A = 1 obs, B = 2 obs, etc.



The following is part of the log file, just to give you a flavour (The rest is on the web site in example1.log)

NOTE: Copyright (c) 1999-2001 by SAS Institute Inc., Cary, NC, USA.
NOTE: SAS (r) Proprietary Software Release 8.2 (TS2M0)
Licensed to OHIO STATE UNIVERSITY -ACADEMIC TECHNOLOGY SERVICE,
Site 00163
55002.
NOTE: This session is executing on the SunOS 5.8 platform.

This message is contained in the SAS news file, and is presented upon initialization. Edit the files "news" in the "misc/base" directory to display site-specific news and information in the program log. The command line option "-nonews" will prevent this display.

NOTE: SAS initialization used:
real time 2:12.18
cpu time 1.05 seconds

```
203 * Sample SAS program
204 * Data set is from Dean and Voss (1999) Design and Analysis of
205 * Experiments. Problem 3, page 129.
206
207 * This data set is from an experiment which examined whether there
are
208 * differences in melting times for margarine. Three different
209 * margarines (brands 1 - 3) were studied and butter was also used as
a
210 * control group (brand 4).;
211
212 options linesize=75; /* set the output width to 75 characters */
213
214 data temp;
215     infile 'p147.3';
216     input brand time;
217     invtime=1/time;
218
219 * print the data to see if everything is ok;
```

NOTE: The infile 'p147.3' is:
File Name=/home/irwin/SAS/p147.3,
Owner Name=irwin,Group Name=parstaff,
Access Permission=rw-----,
File Size (bytes)=240

NOTE: 40 records were read from the infile 'p147.3'.
The minimum record length was 5.
The maximum record length was 5.

NOTE: The data set WORK.TEMP has 40 observations and 3 variables.

NOTE: DATA statement used:
real time 1.75 seconds

cpu time 0.01 seconds

```
220 proc print data=temp;
221     title 'Margarine Experiment';
222
223 * Run the ANOVA;
```

NOTE: There were 40 observations read from the data set WORK.TEMP.

NOTE: PROCEDURE PRINT used:

real time 0.61 seconds
cpu time 0.01 seconds

```
224 proc glm;
225     classes brand;
226     model invtime = brand;
227     estimate 'marg vs but' brand 1 1 1 -3/divisor=3;
228     output out=resids predicted=pred residual=z;
229 run;
```

230

```
231 * Switch from data file temp to data file resids;
```

NOTE: The data set WORK.RESIDS has 40 observations and 5 variables.

NOTE: PROCEDURE GLM used:

real time 1.32 seconds
cpu time 0.09 seconds

```
232 data;
233     set resids;
234
235 * Standardize the residuals to have standard deviation 1;
```

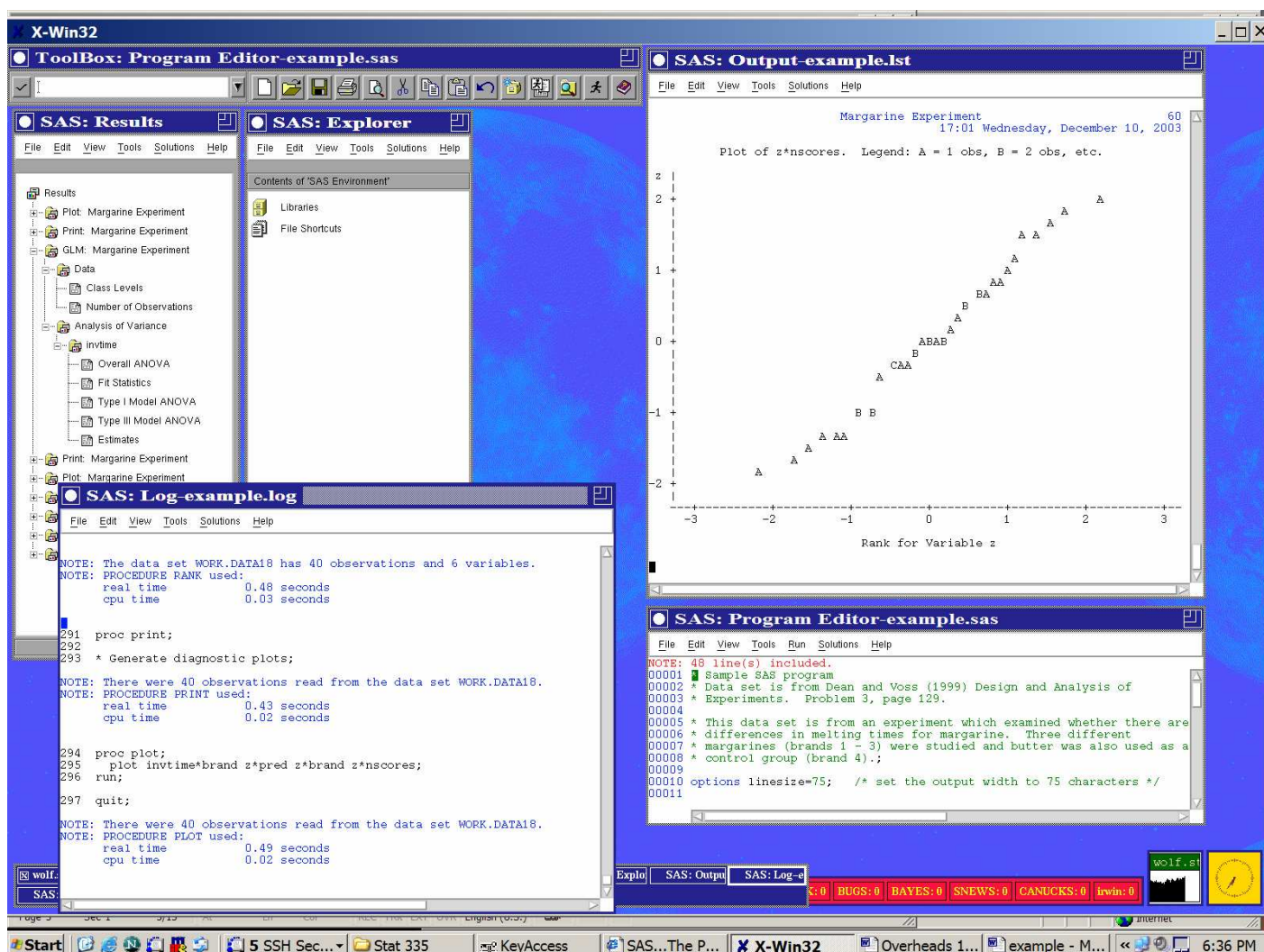
NOTE: There were 40 observations read from the data set WORK.RESIDS.

NOTE: The data set WORK.DATA13 has 40 observations and 5 variables.

NOTE: DATA statement used:

real time 0.69 seconds
cpu time 0.01 seconds

When running the program, the default is to open the windowing environment. The following is from a run on a Sun under unix (Solaris 5.8 under Xwindows). Things will look different on different platforms, but the same set of windows should be available.



SAS can also be run in batch mode (at least with unix and other mainframe operating systems).

To do it in unix, give the command

`sas program.sas`

This will create the files `program.lst` (output) and `program.log` (log file)

SAS System Options

There are many options you can use to alter the way that SAS runs the following are useful for setting up the way SAS generates its output. In the previous output I showed you, I used the defaults, except for `linesize`, which I switched from the default 109.

`CENTER` | `NOCENTER`: Output is either centered or left justified

`DATE` | `NODATE`: Should today's date appear at the top of each page

`NUMBER` | `NONUMBER`: Should the output pages be numbers

`LINESIZE` = *n*: The maximum width of output lines (range = 64 to 256)

`PAGESIZE` = *n*: The maximum number of lines per page of output (range = 15 to 32767)

`PAGENO` = *n*: Starts numbering output pages at zero

`YEARCUTOFF` = *yyyy*: Specifies the first year in a hundred-year span for interpreting two-digit dates.

So to left justify your output and set the pagesize to 50, you can use the command

```
options nocenter pagesize=50;
```

or you can split it across multiple commands

```
options nocenter;
```

```
options pagesize=50;
```

Data Entry

SAS can read in data file from a wide range of formats, text files (space, comma, or tab delimited), Excel, Access, DBF, Lotus 1-2-3, etc.

Two approaches to reading data, DATA step, or PROC IMPORT.

DATA step:

Can be used for a wide range of text files. Often used with space delimited files, but can be used with fixed format files and files with other delimiters. The example earlier was with a space delimited file.

Also I believe that variable names cannot be included in the file. They are given with the `input` part of the DATA step

Now suppose that the data set for the example had brand in columns 1 – 3 and time was in columns 4 - 8. The data could also have been read in with

```
data temp;
  infile 'p147.3';
  input brand 1-3 time 4-8;
  invtime=1/time;
```

You can also a combination of this. For example

```
data temp;
  infile 'p147.3';
  input brand 1-3 time;
  invtime=1/time;
```

would get brand from columns 1-3, and then start looking for time in column 4, but with no restriction on where it ends.

To use a different delimiter, say a comma, modify your `infile` command like

```
infile 'p147.3' dlm = ','
```

For tab delimited files, use `dlm = '09'x` (the tab character is character 9 in ASCII). The `x` indicates that the character is being indicated with a hex code.

In the input line, string variables must be indicated by a `$`. For example

```
input country $ GPA government $
```

would make `country` and `government` string variables and `GPA` numeric.

PROC IMPORT:

This is the approach for reading in non text files and text files with the variable names in the first row. An example is as follows

```
proc import datafile = 'p1473.txt'  
out = temp dbms = tab replace;  
run;
```

```
data;  
  set temp;  
  invtime=1/time;  
run;
```

In this set of commands, the file `p1473.txt` is read in, a data set `temp` is created. The `dbms` indicates that a tab delimited text file is being read in. The `replace` option says to overwrite the data set `temp` if it exists.

Usefile options for `dbms` are `dlim` (space delimited), `cvs` (comma delimited), `excel`, `dbf` (dBase).

The `dbms` option is not needed if the datafile name has the standard file extension in the name.

The data step after it creates a new data file, using a default name (probably `data1`), reading in the data set `temp` created in `proc import` and adds the variable `invtime`.

Writing out data files

This can either be done with a DATA step or PROC export.

DATA step:

This is one approach to writing out your data as a text file. An example is the following

```
data _NULL_;  
  set finaldata;  
  file 'DV.dat';  
  put brand time invtime pred z nscores;  
run;
```

This will write the variables into the file DV.dat, with each variable separated by a space. Other delimiters can be use the dlm option.

The DATA step can also be to create a SAS binary data file.

For example the command

```
data 'DV2';  
  set finaldata;  
run;
```

creates a data file DV2.sas7bdat. The quotes indicate that this should be a permanent file and not a temporary data set. It can be accessed with

```
data test2;  
  set 'DV2';  
  drop nscores;
```

The `drop nscores` piece says to drop that variable from the data set.

```
keep brand time invtime
```

would only keep these three variables and drop the rest.

```
PROC EXPORT:
```

It's basically the opposite of PROC IMPORT, taking most of the same options. The main differences are `outfile` is used instead of `datafile` and `data` is used instead of `out`.

```
proc import outfile = 'p1473.txt'  
data = finaldata dbms = tab;
```

Most of the time, something like this will be used.

Useful PROCS

UNIVARIATE (BASE): Univariate summary statistics

CORR (BASE): Correlations

MEANS (BASE): Similar to Univariate, but adds in some simple testing as well.

FREQ (BASE): Creates and analyzes contingency tables

TABULATE (BASE): Descriptive statistics in tabular format

TTEST (STAT): 1 and 2 sample *t* tests

ANOVA (STAT): Analysis of variance for balanced designs

REG (STAT): Regression

GLM (STAT): General linear model. Similar to `lm()` in S-Plus/R,

LOGISTIC (STAT): Logistic Regression

Common struction for SAS PROCS

```
PROC whatever <options>;  
command1;  
command2;  
...  
run;
```

Options will usually include DATA=. Also common are options of the form OUT=. Note the actual name can change and there may be more than one.

Common commands for many PROC are by, class, freq, output, var, and weight. For example, here are the possible commands that are part of PROC UNIVARIATE:

```
PROC UNIVARIATE <option(s)>;  
  BY <DESCENDING> variable-1  
  <...<DESCENDING> variable-n>  
  <NOTSORTED>;  
  
  CLASS variable-1<(variable-option(s))>  
  <variable-2<(variable-option(s))>>  
  </ KEYLEVEL='value1' | ('value1' 'value2')>;  
  
  FREQ variable;  
  
  HISTOGRAM <variable(s)> </ option(s)>;  
  
  ID variable(s);  
  
  INSET <keyword(s) DATA=SAS-data-set> </
```

```
option(s)>;  
OUTPUT <OUT=SAS-data-set> statistic-  
keyword-1=name(s)  
<... statistic-keyword-n=name(s)>  
<percentiles-specification>;  
PROBPLOT <variable(s)> </ option(s)>;  
QQPLOT <variable(s)> </ option(s)>;  
VAR variable(s);  
WEIGHT variable;
```

The number of options for this PROC is huge. Go to the documentation on the web to see them all.

PROC UNIVARIATE example:

```
options linesize=75;
```

```
DATA swiss;
```

```
  infile '/home/irwin/SAS/swiss.dat';
```

```
  input fertility agriculture examination
```

```
    education catholic infantmort;
```

```
  title 'Swiss Data Set';
```

```
proc univariate data = swiss
```

```
  normal
```

```
  cibasic
```

```
  plots
```

```
  alpha = 0.01;
```

```
  var education;
```

```
run;
```

And the output is

Swiss Data Set 1
 13:26 Thursday, December 11, 2003

The UNIVARIATE Procedure
 Variable: education

Moments

N	47	Sum Weights	47
Mean	10.9787234	Sum Observations	516
Std Deviation	9.61540739	Variance	92.4560592
Skewness	2.42076401	Kurtosis	7.43408377
Uncorrected SS	9918	Corrected SS	4252.97872
Coeff Variation	87.5821991	Std Error Mean	1.40255132

Basic Statistical Measures

Location		Variability	
Mean	10.97872	Std Deviation	9.61541
Median	8.00000	Variance	92.45606
Mode	7.00000	Range	52.00000
		Interquartile Range	6.00000

Basic Confidence Limits Assuming Normality

Parameter	Estimate	99% Confidence Limits	
Mean	10.97872	7.21005	14.74740
Std Deviation	9.61541	7.55881	13.03220
Variance	92.45606	57.13564	169.83834

Tests for Location: Mu0=0

Test	-Statistic-	-----p Value-----	
Student's t	t 7.82768	Pr > t	<.0001
Sign	M 23.5	Pr >= M	<.0001
Signed Rank	S 564	Pr >= S	<.0001

Tests for Normality

Test	--Statistic--	-----p Value-----	
Shapiro-Wilk	W 0.748203	Pr < W	<0.0001
Kolmogorov-Smirnov	D 0.246538	Pr > D	<0.0100
Cramer-von Mises	W-Sq 0.656644	Pr > W-Sq	<0.0050
Anderson-Darling	A-Sq 3.587264	Pr > A-Sq	<0.0050

Quantiles (Definition 5)

Quantile	Estimate
100% Max	53
99%	53

The UNIVARIATE Procedure
 Variable: education

Quantiles (Definition 5)

Quantile	Estimate
95%	29
90%	28
75% Q3	12
50% Median	8
25% Q1	6
10%	3
5%	2
1%	1
0% Min	1

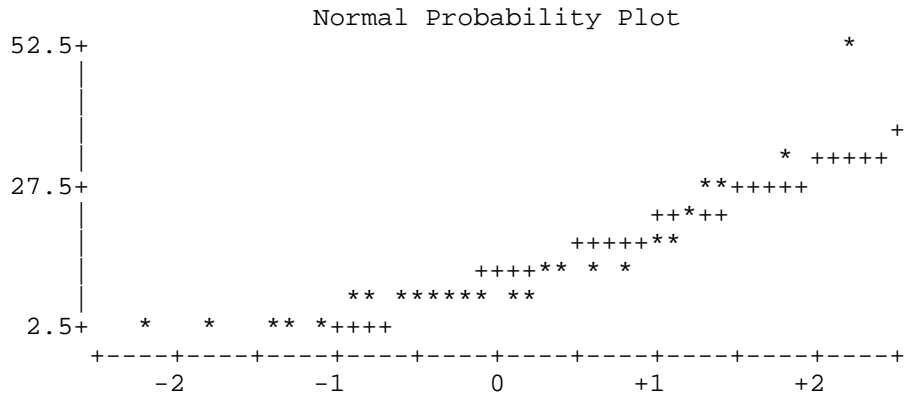
Extreme Observations

----Lowest----		----Highest---	
Value	Obs	Value	Obs
1	25	28	18
2	33	29	46
2	31	29	47
2	16	32	42
3	37	53	45

Stem	Leaf	#	Boxplot
5	3	1	*
4			
4			
3			
3	2	1	*
2	899	3	0
2	0	1	
1	59	2	
1	00122222333	11	+---+---+
0	5566667777778888999	20	*-----*
0	12223333	8	
-----+-----+-----+-----+			

Multiply Stem.Leaf by 10**+1

The UNIVARIATE Procedure
Variable: education



PROC GLM:

Has a similar function to `lm()` in S-Plus/R.

```
PROC GLM < options > ;  
  CLASS variables ;  
  MODEL dependents=independents < / options > ;  
  ABSORB variables ;  
  BY variables ;  
  FREQ variable ;  
  ID variables ;  
  WEIGHT variable ;  
  CONTRAST 'label' effect values < ... effect values > < / options > ;  
  ESTIMATE 'label' effect values < ... effect values > < / options > ;  
  LSMEANS effects < / options > ;  
  MANOVA < test-options >< / detail-options > ;  
  MEANS effects < / options > ;  
  OUTPUT < OUT=SAS-data-set >  
    keyword=names < ... keyword=names > < / option > ;  
  RANDOM effects < / options > ;  
  REPEATED factor-specification < / options > ;  
  TEST < H=effects > E=effect < / options > ;
```

```

options linesize=75;

proc import datafile = '93cars.txt'
    out = cars;
run;

data cars93;
    set cars;
    highfuel = 100/highmpg;
    cityfuel = 100/citympg;
run;

proc glm data = cars93;
    classes domestic;
    model highfuel = weight domestic
        weight*domestic / solution clm;
    output out=resids predicted=pred
        residual=resids;
run;

```

In the model statement, the structure is similar to S-Plus/R, in that interactions are denoted with *. However when an interaction term is included, the main effect are not automatically included. They must be explicitly added as I did above.

Also it takes a wide range of options. For example you must explicitly ask for the parameter estimates.

The classes statement indicates which variables in the model should be treated as categorical factors. The default is to treat as variables in the model as numeric.

When fitting models with categorical variables, SAS deals with the parametrization and contrasts a bit differently. It actually fits an over parametrized model and uses a generalized inverse instead of a regular inverse, since its $X^T X$ matrix is singular.

There is a wide range of summary statistics that can be output. They include, the fits, residuals, leverages, Cook's D, prediction intervals and confidence intervals, DFFITS, etc.

The SAS System 1
 14:09 Thursday, December 11, 2003

The GLM Procedure

Class Level Information

Class	Levels	Values
Domestic	2	0 1

Number of observations 93

The SAS System 2
 14:09 Thursday, December 11, 2003

The GLM Procedure

Dependent Variable: highfuel

Source	DF	Sum of Squares	Mean Square	F Value	Pr > F
Model	3	22.51693938	7.50564646	69.45	<.0001
Error	89	9.61827575	0.10807051		
Corrected Total	92	32.13521513			

R-Square Coeff Var Root MSE highfuel Mean
 0.700694 9.282533 0.328741 3.541499

Source	DF	Type I SS	Mean Square	F Value	Pr > F
Weight	1	22.00269643	22.00269643	203.60	<.0001
Domestic	1	0.02634815	0.02634815	0.24	0.6227
Weight*Domestic	1	0.48789481	0.48789481	4.51	0.0364

Source	DF	Type III SS	Mean Square	F Value	Pr > F
Weight	1	21.19245842	21.19245842	196.10	<.0001
Domestic	1	0.42981420	0.42981420	3.98	0.0492
Weight*Domestic	1	0.48789481	0.48789481	4.51	0.0364

Parameter	Estimate	Standard Error	t Value	Pr > t
Intercept	1.368412490 B	0.27519967	4.97	<.0001
Weight	0.000706864 B	0.00008484	8.33	<.0001
Domestic 0	-0.742054400 B	0.37209083	-1.99	0.0492
Domestic 1	0.000000000 B	.	.	.
Weight*Domestic 0	0.000252874 B	0.00011901	2.12	0.0364
Weight*Domestic 1	0.000000000 B	.	.	.

NOTE: The X'X matrix has been found to be singular, and a generalized inverse was used to solve the normal equations. Terms whose estimates are followed by the letter 'B' are not uniquely estimable.

The GLM Procedure

Observation	Observed	Predicted	Residual
1	3.22580645	3.22245051	0.00335595
2	4.00000000	4.04302685	-0.04302685
3	3.84615385	3.86547524	-0.01932140
4	3.84615385	3.89426740	-0.04811355
5	3.33333333	4.11980592	-0.78647259
6	3.22580645	3.40418148	-0.17837502
57	4.00000000	3.40480080	0.59519920
58	3.44827586	3.42879426	0.01948160
59	4.00000000	4.00943601	-0.00943601
60	3.84615385	3.10022986	0.74592399

Observation	95% Confidence Limits for Mean Predicted Value	
1	3.11742228	3.32747873
2	3.90169213	4.18436157
3	3.74451763	3.98643286
4	3.77029337	4.01824142
5	3.96857802	4.27103383
6	3.29594990	3.51241305
57	3.30711142	3.50249018
58	3.33135039	3.52623814
59	3.87225027	4.14662174
60	2.94315247	3.25730724

Observation	Observed	Predicted	Residual
61	3.84615385	3.92019237	-0.07403852
62	3.03030303	2.82895775	0.20134528
63	4.16666667	4.20618238	-0.03951572
91	4.00000000	3.32322304	0.67677696
92	3.57142857	3.49117726	0.08025131
93	3.57142857	3.74070925	-0.16928068

Observation	95% Confidence Limits for Mean Predicted Value	
61	3.80282367	4.03756106
62	2.68401721	2.97389830
63	4.04325020	4.36911457
91	3.22340681	3.42303927
92	3.39354703	3.58880749
93	3.63115860	3.85025990

The GLM Procedure

Sum of Residuals	-0.00000000
Sum of Squared Residuals	9.61827575
Sum of Squared Residuals - Error SS	-0.00000000
PRESS Statistic	10.56191586
First Order Autocorrelation	0.07667443
Durbin-Watson D	1.84367064

Here is the R output for the same model

```
> anova(weight.domestic.int.lm)
Analysis of Variance Table
```

```
Response: Highfuel
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
Weight	1	22.0027	22.0027	203.5957	< 2e-16	***
Domestic	1	0.0263	0.0263	0.2438	0.62269	
Weight:Domestic	1	0.4879	0.4879	4.5146	0.03638	*
Residuals	89	9.6183	0.1081			

```
> summary(weight.domestic.int.lm)
```

```
Call:
```

```
lm(formula = Highfuel ~ Weight * Domestic, data = cars93)
```

```
Residuals:
```

Min	1Q	Median	3Q	Max
-0.78647	-0.21346	-0.03952	0.17163	0.99145

```
Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	6.264e-01	2.504e-01	2.501	0.0142	*
Weight	9.597e-04	8.347e-05	11.498	<2e-16	***
DomesticDomestic	7.421e-01	3.721e-01	1.994	0.0492	*
Weight:DomesticDomes	-2.529e-04	1.190e-04	-2.125	0.0364	*

Notice that the summaries are similar. Note that R generates the Type I (Sequential SS)