

## SAS Advanced Programming with Efficiency in Mind: A Real Case Study

Lingqun Liu, University of Michigan, Ann Arbor, MI

### ABSTRACT

This paper uses a real work example to demonstrate the concept and some basic tips of SAS programming efficiency. The first section of the paper introduces the background of a SAS application and its performance metrics. The second section analyzes the structure and features of the SAS application. The third section analyzes the log of the application to identify efficiency issues. In addition, in this section a log analysis utility is introduced. The fourth section provides a re-developed version of the application with performance improved to reduce 99.6% of its runtime. The last section tries to raise awareness of SAS programming efficiency and suggests some basic tips. The application discussed in the paper has been tested with SAS 9.2, 9.3 and 9.4 on Windows machines. The target audience includes SAS programmers from beginner to advanced level.

### INTRODUCTION

Did you ever have any SAS applications that took longer than you expected to run? It could be hours, or even days long. Most of us find it frustrating when things like that happened, especially when you had a tight deadline to meet, or you had to run the job many times within a limit of time. Many programmers might think it is caused by the nature of their SAS application, such as big data sets, complex process, and limitation of computing power and resources, etc. Moreover, it was not uncommon that many SAS application developers/programmers did not realize that there usually were efficiency issues. This paper uses an example to raise the awareness of SAS programming efficiency, introduce a log analysis utility, and provide some basic tips.

### I.1 BACKGROUND

UM-KECC is a multidisciplinary research center within the UM School of Public Health (SPH). UM-KECC was formed in 1993 and its mission is “to promote health, improve clinical practice and patient outcomes, optimize resource utilization, and inform public policy regarding organ failure and organ transplantation.” UM-KECC pursues this mission “through high quality research, advances in biostatistics, and post-graduate education and training.” ([www.kecc.sph.umich.edu](http://www.kecc.sph.umich.edu)).

UM-KECC has been working with CMS to develop quality measures of ESRD patient care for years. Each quarter, as one tiny part of the large efforts, UM-KECC produces lists of ESRD patients included in the dialysis facility compare (QDFC) measures for more than 6,000 Medicare dialysis facilities nationwide. There are five measures: M1, M2, M3, M4, and M5. In each quarter, there are more than 21.7K patient list files (21,870 for 201607, 21,702 for 201604) created. This whole process consists of five similar SAS jobs, one for each measure.

M1\_DFC\_Patient\_Lists.sas  
M2\_DFC\_Patient\_Lists.sas  
M3\_DFC\_Patient\_Lists.sas  
M4\_DFC\_Patient\_Lists.sas  
M5\_DFC\_Patient\_Lists.sas

## I.2 PROCESS TIME

The process time varies for the jobs. The M5 job took about 10 hours. (And it could occasionally even take longer than 69 hours for some reason in reality. It was the worst case we had!) The rest took from 18 seconds to around 16 hours. The total process times for the last two quarters were about 19.4 hours and 33.4 hours.

Jobs	201604		201607	
	Real time	CPU time	Real time	CPU time
M1_DFC_Patient_Lists.sas	4:16:02.04	4:04:10.54	4:11:53.18	4:04:35.88
M2_DFC_Patient_Lists.sas	16:30.83	13:40.53	1:48:09.24	15:27.02
M3_DFC_Patient_Lists.sas	1:39.45	18.93	2:30.09	22.88
M4_DFC_Patient_Lists.sas	4:49:30.74	4:42:13.26	16:17:07.14	7:34:12.81
M5_DFC_Patient_Lists.sas	10:02:30.96	9:49:39.17	11:02:17.87	10:22:32.44
total	<b>19:26:14.02</b>	17:08:55.50	<b>33:21:57.52</b>	22:17:11.03

## II. CODE ANALYSIS

You may wonder why some of these simple jobs can take more than 10 hours. Moreover, 19 to 33 hours of total runtime of the production is way too long. Are there any efficiency issues? Can the application be improved? Let us start with examining the SAS code, in order to see what the issues could be and identify how to fix them. In the following sections, our analysis and redevelopment will use M5 job as an example. The rest of these jobs are identical in terms of the code design, structure, functionality, and issues, etc. Please see Figure 1.1 and Figure 1.2 for the code listings.

Figure 1.1 Original Code Snapshot One

```

25
26 proc sort data=faclib.facinfo_&lookupdtd. out= facinfo (keep= facid network provname provcity state)
27 where DFC_report=1;
28 by facid;
29 run;
30
31 %macro print_list(data,measure,name);
32
33 %if &measure=M5 %then %do;
34 %let vars=firsts dialysis_90days age_ge_18 calcium_uncorrected in_facility modality elig_pm avg_3mo
35 %let varslabel= firsts='First'service^date' dialysis_90_days='Dialysis^ge^90^days' age_ge_18='Patie
36 in_facility='Meets^facility^requirement' modality='Meets^modality^requirement' elig_pm='Eligible^pa
37 hypercal_gt10_2='Hypercalemia-gt^10.2';
38 %end;
39
40 %put &vars;
41 %put &varslabel;
42
43 *****
44 -----Sort measure files, keep only variables output to list-----
45 *****
46
47 proc sort data=mlib.&data. out=temp(keep=patid facid &vars year month quarter);
48 by patid;
49 run;
50
51 /*****
52 *---Merge individual measure files with patients to get patient identifiers---*
53 /*****
54 data saflib.M5_plist_&dateit.;
55 merge temp (in=a) saflib.patients (keep=patid surname first_name m_initial ssn);
56 by patid;
57 if a;
58 fname=trim(first_name)||' '||trim(m_initial);
59 Patient_id_n_;
60 ssn1=SSN+0;
61 **** Note: In this step, a small percentage of pts have characters in their SSN. This ****
62 * causes warning messages in the log file because ssn1 cannot be calculated, and in the *
63 * final patient list they will have a missing SSN. Since the SSNs are not numeric, we *
64 * assume they are not valid, so having missing SSN is not a problem. *
65 run;
66
67 /*****
68 *---Merge with facinfo to obtain provider name, city, state, etc---*
69 /*****
70 proc sort data=saflib.M5_plist_&dateit.;
71 by facid;
72 run;
73

```

## CODE LOGIC

It is a simple job and it has two requirements:

1. **Create data:** Put facility information (6,499 observations), patient information (2,819,069 observation) and measure results (6,423,888 observations) together to create a patient-measure level data set containing information for patients included in the measure for all facilities. Also, perform a few data manipulations.
2. **Print data:** Print patient-measure information by facility in plain text format with file extension .txt.

Figure 1.2 Original Code Snapshot Two

```
1\MWSUG16_BB18\MS_DFC_Patient_Lists_original_MWSUG.sas *
78 data &measure._ptlist;
79 merge saflib.M5_plist_&dateit.(in=a) facinfo (in=infacinfo);
80 by facid;
81 if a and infacinfo;
82 facility=trim(provname)||', '||trim(provcity)||', '||state;
83 format ssn1 ssn11.;
84 report_period=strip(year)||' '||strip(month)||' '||strip(quarter);
85 run;
86
87
88 proc sort data= &measure._ptlist;
89 by network facid facility surname first_name ;
90 run;
91
92
93 proc sql;
94 select count(distinct facid) into: numprov;
95 from &measure._ptlist;
96 quit;
97
98 %put &numprov;
99
100 data _null_;
101 length numprovsc $9.;
102 numprovsc=strip(&numprov);
103 call symput('numprovsc', numprovsc);
104 run;
105
106
107 proc sql;
108 select distinct facid into :prov1 -:prov&numprovsc notrim
109 from &measure._ptlist;
110 quit;
111
112
113 %do i=1 %to &numprov;
114 %put '*****' &&prov&i;
115 data prulevel ;
116 set &measure._ptlist;
117 where facid="&&prov&i";
118 call symput ("Facility", compress(facility,""));
119
120 run;
121 %put '*****' &facility;
122
123 ods listing file="&outfile";
124 title "CONFIDENTIAL: Patients included in the &name. measure reported in the";
125 title2 "Quarterly Dialysis Compare-Preview for &month., &year. report.";
126 title3 "MMH Certification Number=&&prov&i Facility=&facility";
127 options ls=max ps=85;
128 proc print data=prulevel noobs split='^' uniform;
```

## CODE DESIGN AND STRUCTURE

This code has two parts, one for each subtask. The first part consists of PROCs and DATA steps. The second subtask is implemented with a %MACRO %do loop that creates and prints out one data set for each facility. As a result, there are more than 6,000 DATA steps and PROCs generated by the %MACRO/%DO loop at runtime.

1. **Create data:** Four PROC SORTs, two DATA MERGES.
2. **Print data:** Two PROC SQLs, one DATA \_NULL\_, one %MACRO %do loop of 1 DATA step and ODS/PROC PRINT.

## SAS FEATURES

There are many SAS features, including some advanced ones, in this SAS application.

- DATA STEP MERGE, PROC SQL, PROC SORT;
- %MACRO, &&VAR&N, CALL SYMPUT, INTO:, %Do loop; DATA \_NULL\_;
- Data type conversion (+0), function COMPRESS(), STRIP(), TRIM();
- ODS LISTING, Dynamic titles, PROC PRINT options, etc.
- System options: LS, NODATE, NONUMBER, NOCENTER, ERRORS, SOURCE2, MPRINT.

## CRITICAL THINKING

Does it need to be so complicated (using so many steps and features)? Is %macro really needed? (Can the %macro be avoided?) Which features/steps did take most of the runtime? Would the large number of small DATA steps and PROCs be an efficiency issue? Or is the long runtime due to the large size of the input SAS data sets? To answer these questions, I inspected the log files of the job along with the SAS code.

## III.1 LOG ANALYSIS: OBSERVATION & ESTIMATION

The log file is lengthy. It has more than 45,000 lines. We need to search for the key words ‘real time’ to see how long each step took. First, let us look at the runtime for task one -- the creation of measure-patient data set. The facility info data has about 6,600 records. The measure data has about 6.5 million observations. The patient info data set has about 2.5 million records. The DATA step and PROC SORT processed these data sets within a few minutes. It is fast to create the measure-patient data set. Since SAS is so powerful, the sizes of the data sets in this application are not the issue (Please see Figure 2.1 and Figure 2.2 for details.)

Figure 2.1 Log Snapshot One

```
I:\MWSUG16_BB18\M5_DFC_Patient_Lists_MWSUG (2).log
461 NOTE: There were 2819069 observations read from the data set SAFKECC.PATIENTS.
462 NOTE: The data set SAFLIB.M5_PLIST_201604 has 6423888 observations and 21 variables.
463 NOTE: Compressing data set SAFLIB.M5_PLIST_201604 decreased size by 42.01 percent.
464 Compressed is 64233 pages; un-compressed would require 110757 pages.
465 NOTE: DATA statement used (Total process time):
466 real time 1:33.79
467 cpu time 37.01 seconds
468
469 |
470 SYMBOLGEN: Macro variable DATEIT resolves to 201604
471 MPRINT(PRINT_LIST): proc sort data=SAFLIB.M5_plist_201604;
472 MPRINT(PRINT_LIST): by provfs;
473 MPRINT(PRINT_LIST): run;
474
475 NOTE: There were 6423888 observations read from the data set SAFLIB.M5_PLIST_201604.
476 NOTE: The data set SAFLIB.M5_PLIST_201604 has 6423888 observations and 21 variables.
477 NOTE: Compressing data set SAFLIB.M5_PLIST_201604 decreased size by 42.00 percent.
478 Compressed is 64234 pages; un-compressed would require 110757 pages.
479 NOTE: PROCEDURE SORT used (Total process time):
480 real time 1:36.95
481 cpu time 36.65 seconds
482
```

Figure 2.2 Log Snapshot Two

```
I:\MWSUG16_BB18\M5_DFC_Patient_Lists_MWSUG (2).log *
515 NOTE: The data set WORK.M5_PTLIST has 6423888 observations and 27 variables.
516 NOTE: Compressing data set WORK.M5_PTLIST decreased size by 51.77 percent.
517 Compressed is 88522 pages; un-compressed would require 183540 pages.
518 NOTE: PROCEDURE SORT used (Total process time):
519 real time 2:47.55
520 cpu time 1:04.28|
521
```

Now let us look at the runtime for task two – the creation of the facility specific patient list files. Every time a list file was created, one small DATA step and one PROC PRINT were executed. After scanning the log file, we noticed that the process only used about 5.3 seconds or so per facility.

However, since there were more than 6,000 facilities, the total runtime ended up as about 10 hours. The stop value of the %DO loop was 6,375 for this case. Therefore, the total run time was about  $5.28 * 6375 / (60 * 60)$  seconds = 9.35 hours. (Please see Figure 2.3 and Figure 2.4 for details)

Figure 2.3 Log Snapshot Three

```

I:\MWSUG16_BB18\M5_DFC_Patient_Lists_MWSUG (2).log
578 NOTE: There were 1404 observations read from the data set WORK.MBS_PTLIST.
579 WHERE pr= '000000';
580 NOTE: The data set WORK.PRULEVEL has 1404 observations and 27 variables.
581 NOTE: Compressing data set WORK.PRULEVEL decreased size by 53.66 percent.
582 Compressed is 19 pages; un-compressed would require 41 pages.
583 NOTE: DATA statement used (Total process time):
584 real time 5.28 seconds
585 cpu time 5.28 seconds

```

Figure 2.4 Log Snapshot Four

```

I:\MWSUG16_BB18\M5_DFC_Patient_Lists_MWSUG (2).log
457571 MPRINT(PRINT_LIST): ods listing close;
457572 MLOGIC(PRINT_LIST): %DO loop index variable I is now 6376; loop will not iterate again.
457573 MLOGIC(PRINT_LIST): Ending execution.
457574 294
457575 295
457576
457577 NOTE: SAS Institute Inc., SAS Campus Drive, Cary, NC USA 27513-2414
457578 NOTE: The SAS System used:
457579 real time 10:02:30.96
457580 cpu time 9:49:39.17
457581

```

### III.2 LOG ANALYSIS: STATISTICS

To get the statistics of the runtime of the SAS application, I developed a simple SAS utility to analyze the full lengthy SAS log file (457,581 lines in this case). The log analysis utility consists of two small %macros: %log\_io\_search(), %log\_io\_data() and a PROC MEANS. (Please see Figure 3.1 for details.)

Figure 3.1 Log Analysis code

```

Programmer's File Editor - [log_analysis_MWSUG.sas]
File Edit Options Template Execute Macro Window Help
1 *****;
2 *Program name: log_analysis.sas
3 *By : lqliugumich.edu 2008, 2016
4 *Purpose : to process SAS log file to analyze SAS application
5 * : --structure and performance
6 *
7 *Input : SAS log file
8 *Output : two txt file and two datasets, plus ...
9 *
10 *Note : internal use only
11 * : some lines in step 3 need revision to reflect individual needs
12 *****;
13 option mprint;
14 ***** STEP 1 *****;
15 ** usage: -----;
16 ** %log_io_search(log=[your log file].log,
17 doc=[results txt file].txt);
18 *****;
19 %macro log_io_search(log=,doc=);
20 %if not %index(&log, '.') %then %let log=&log.*.log;
21 data _null_;
22 length logname $200 ;
23 infile "&log" filename=f end=done;
24 file "&doc";
25 logname=f;
26 if logname ne lag(logname) then do;
27 if line then put line "lines read";
28 put // '-----' logname '-----';
29 line=0;
30 end;
31 input ;
32 line + 1;
33 output = index(_infile_,'NOTE: The data set') and
34 not index(_infile_,'-- NOTE:');
35 or
36 index(_infile_,'were written to the file');
37 ;
38 input = index(_infile_,'read from') or index(_infile_,'WHERE ') ;
39 time = index(_infile_,' time') ;
40 logline = _infile_;
41 keep= ifc(input,'INPUT ','OUTPUT') ;
42 keep= ifc(input,keep,'TIME ');
43 if input or output or time then put keep logline;
44 if done then put line "lines read";
45 run;
46 %mend;
47
48 %let log=I:\MWSUG16_BB18\M5_DFC_Patient_Lists_MWSUG.log;
49 %let doc=I:\MWSUG16_BB18\M5_DFC_Patient_Lists_MWSUG.txt;
50 %log_io_search(log=&log,doc=&doc);

```

```

47 ;
48 input = index(_infile_,'read from') or index(_infile_,'WHERE ');
49 time = index(_infile_,'time');
40 logline = _infile_;
41 keep= ifc(input,'INPUT ','OUTPUT ');
42 keep= ifc(input,keep,'TIME ');
43 if input or output or time then put keep logline;
44 if done then put line "Lines read";
45 run;
46 %mend;
47 |
48 %let log=I:\MWSUG16_BB18\M5_DFC_Patient_Lists_MWSUG.log;
49 %let doc=I:\MWSUG16_BB18\M5_DFC_Patient_Lists_MWSUG.txt;
50 %log_io_search(log=&log,doc=&doc);
51
52 ***** STEP 2*****;
53 ** usage: -----;
54 ** %log_io_data([results txt file from step 1 above].txt,
55               doc=[results txt file].txt);
56 *****;
57 %macro log_io_data(log=&log);
58 data log_runtime_nessy log_runtime(keep= dsn ntime ctime procdat obs);
59 length logname $ logline $200 dsn $32 procdat $6;
60 retain dsn obs;
61 infile "&log" filename=f end=done;
62 file "&doc"; *optional;
63 logname=f;
64 if logname ne lag(logname) then do;
65   if line then put line "lines read";
66   put // "-----" logname "-----";
67   line=0;
68 end;
69 input @;
70 if
71   index(_infile_,'TIME NOTE: The data set')
72   or index(_infile_,'TIME NOTE: DATA statement used (Total process time):')
73   or index(_infile_,'TIME NOTE: PROCEDURE SORT used (Total process time):')
74   or index(_infile_,'TIME NOTE: PROCEDURE SQL used (Total process time):')
75 then
76

```

The first macro %log\_io\_search() uses a DATA\_NULL\_ step to search through the log file, extract the key information for each step, and write out them into a txt file. (Figure 3.2)

Figure 3.2 Log Analysis Results Snapshot One

```

I:\MWSUG16_BB18\M5_DFC_Patient_Lists_MWSUG.txt
|
-----I:\MWSUG16_BB18\M5_DFC_Patient_Lists_MWSUG.log -----
TIME real time          0.10 seconds
TIME cpu time           0.06 seconds
INPUT NOTE: There were 6553 observations read from the data set FACLIB.FACINFO_201601.
INPUT WHERE DFC_report=1;
TIME NOTE: The data set WORK.FACINFO has 6553 observations and 5 variables.
TIME NOTE: PROCEDURE SORT used (Total process time):
TIME real time          0.74 seconds
TIME cpu time           0.06 seconds
INPUT NOTE: There were 6554484 observations read from the data set MLIB.M5_PATIENT_LIST.
TIME NOTE: The data set WORK.TEMP has 6554484 observations and 14 variables.
TIME NOTE: PROCEDURE SORT used (Total process time):
TIME real time         19.53 seconds
TIME cpu time          18.54 seconds

```

The second macro %log\_io\_data() again uses a DATA step to search through the output text file generated from the first step and put the results in a better text format. (Figure 3.3)

Figure 3.3 Log Analysis Results Snapshot Two

```

I:\MWSUG16_BB18\M5_DFC_Patient_Lists_MWSUG2.txt
|
-----I:\MWSUG16_BB18\M5_DFC_Patient_Lists_MWSUG.txt -----
SORT: TIME real time          0.74 seconds          6,553          WORK.FACINFO ---ntime=0:00.74
SORT: TIME real time         19.53 seconds        6,554,484          WORK.TEMP ---ntime=0:19.53
DATA: TIME real time         27.00 seconds        6,554,484          SAFLIB.M5_PLIST_201607 ---ntime=0:27.00
SORT: TIME real time         18.59 seconds        6,554,484          SAFLIB.M5_PLIST_201607 ---ntime=0:18.59
DATA: TIME real time         26.34 seconds        6,554,484          WORK.M5_PTLIST ---ntime=0:26.34
SORT: TIME real time         29.15 seconds        6,554,484          WORK.M5_PTLIST ---ntime=0:29.15
SQL : TIME real time          7.20 seconds        6,554,484          WORK.M5_PTLIST ---ntime=0:07.20
DATA: TIME real time          0.00 seconds        6,554,484          WORK.M5_PTLIST ---ntime=0:00.00
SQL : TIME real time          6.75 seconds        6,554,484          WORK.M5_PTLIST ---ntime=0:06.75
DATA: TIME real time          5.21 seconds          1,380          WORK.PRULEVEL ---ntime=0:05.21
PRNT: TIME real time          0.01 seconds          1,380          WORK.PRULEVEL ---ntime=0:00.01
DATA: TIME real time          5.21 seconds          720          WORK.PRULEVEL ---ntime=0:05.21
PRNT: TIME real time          0.00 seconds          720          WORK.PRULEVEL ---ntime=0:00.00

```

In addition, it puts them into a SAS data set for further analysis. (Figure 3.4)

**Figure 3.4 Log Analysis Results Snapshot Three**

	dsn	PROCDAT	obs	ctime	ntime
1	WORK.FACINFO	SORT:	6553	0.74	0.74
2	WORK.TEMP	SORT:	6554484	19.53	19.53
3	SAFLIB.M5_PLIST_201607	DATA:	6554484	27.00	27
4	SAFLIB.M5_PLIST_201607	SORT:	6554484	18.59	18.59
5	WORK.M5_PTLIST	DATA:	6554484	26.34	26.34
6	WORK.M5_PTLIST	SORT:	6554484	29.15	29.15
7	WORK.M5_PTLIST	SQL :	6554484	7.20	7.2
8	WORK.M5_PTLIST	DATA:	6554484	0.00	0
9	WORK.M5_PTLIST	SQL :	6554484	6.75	6.75

Then PROC MEANS summarizes the runtime of the whole process recorded in the SAS log file. As an example, the statistics of the M5 job for the 201607 run are shown below (Figure 3.5).

**Figure 3.5 Log Analysis Results**

PROCDAT	dsn	N Obs	Variable	Mean	Maximum	Minimum	Sum
DATA:	SAFLIB.M5_PLIST_201607	1	ntime	27.0000000	27.0000000	27.0000000	27.0000000
		obs		6554484.00	6554484.00	6554484.00	6554484.00
	WORK.M5_PTLIST	2	ntime	13.1700000	26.3400000	0	26.3400000
		obs		6554484.00	6554484.00	6554484.00	13108968.00
	WORK.PRLEVEL	6426	ntime	6.1020090	20.9000000	5.1600000	39211.51
		obs		1019.99	6240.00	12.0000000	6554484.00
PRNT:	WORK.PRLEVEL	6426	ntime	0.0066355	0.0700000	0	42.6400000
		obs		1019.99	6240.00	12.0000000	6554484.00
SORT:	SAFLIB.M5_PLIST_201607	1	ntime	18.5900000	18.5900000	18.5900000	18.5900000
		obs		6554484.00	6554484.00	6554484.00	6554484.00
	WORK.FACINFO	1	ntime	0.7400000	0.7400000	0.7400000	0.7400000
		obs		6553.00	6553.00	6553.00	6553.00
	WORK.M5_PTLIST	1	ntime	29.1500000	29.1500000	29.1500000	29.1500000
		obs		6554484.00	6554484.00	6554484.00	6554484.00
	WORK.TEMP	1	ntime	19.5300000	19.5300000	19.5300000	19.5300000
		obs		6554484.00	6554484.00	6554484.00	6554484.00
SQL :	WORK.M5_PTLIST	2	ntime	6.9750000	7.2000000	6.7500000	13.9500000
		obs		6554484.00	6554484.00	6554484.00	13108968.00

There are 6,428 DATA steps, 3 large ones, and 6,426 small ones. The large data steps only took a few minutes. And the 6,426 small data steps took more than 10 hours: 39,211/(60\*60) seconds = 10.89 hours. The PROC steps took less than a minute.

Based on the statistics shown above, we can tell that the %MACRO/%DO structure is very time consuming in this application. It posts an efficiency issue. In the next section, we will show the redevelopment of this application to make it more efficient.

#### IV. REDEVELOPING THE APPLICATION

Once we have identified the cause of the long runtime, we can redesign the application with efficiency in mind.

The first area to improve the original SAS application is to reduce the number of steps. Some data steps and procs can be combined, some steps and the %macro and data sorting can be avoided. SAS view can be used to replace data set. In addition, we can reduce the size of the log file by getting rid of macro related lines and fixing invalid data errors. That will make the log file more readable and save some I/O time as well. Second, and most importantly, for the reporting part, we can use a simple but powerful technique to avoid the 6,000+ small data steps: We use the SAS BY processing mechanism and DATA step **FILE statement** instead of the loops of DATA steps and PROC PRINTs.

Here is the outline of the re-developed SAS application. The new code only contains one PROC SQL view and one DATA step. There is no %macro/PROC PRINT/SORT. It uses a DATA step FILE statement with option FILEVAR= to write out facility specific reports.

```
PROC SQL; CREATE VIEW ... AS ...; QUIT;

DATA ...;
  SET ...;
  BY FACID;
  ...
  _FN= ... FACID ...;
  FILE WRITEOUT FILEVAR=_FN ...;
  ...
  PUT ...;
  ...
RUN;
```

The new SAS application has only about 80 lines. (The original one has about 150 lines.)

Figure 4.1a Redeveloped Code (part 1)

```
1 /*****
2 Program Name: M5_PatLis.sas
3 Purpose      : Print facility patient list for M5 measure for DFC
4 By          : lqliu@umich.edu 2016-04-25
5
6 Input       : 1. measure results- QDFC.M5_patient_list
7             : 2. patient info - saflib.patients
8             : 3. facility info - faclib.facinfo_lookupdt
9
10 Output      : 1. facility patient lists: &outpath\M5_PatList_999999.lst
11             : 2. SAS dataset -- saflib.M5_plist
12
13 Note       : A programmer with appropriate permissions must run this code
14             : Make sure output folder has been created.
15
16 *****/
17
18 %include "\\disk\DFC\Code\dfc_dateparms.sas";
19 options ls=max ps=85 nodate nonumber source;
20
21 *--Output for patient lists--*
22 libname ptlists "\\disk\QDFC\Patient_Lists\&refreshdt._release&dateit.";
23 libname saflib "\\disk\saflib";
24
25 %let outpath=\\disk\QDFC\Patient_Lists\&refreshdt._release&dateit.;
26 %let outpath=\\disk\QDFC\Patient_Lists\quinn_test_output; * testing path;
27 %let runby=quinn; * for testing;
28
29 %let vars=firsts dialysis_90days age_ge_18 calcium_uncorrected in_facility modality elig_pm
30
31 *-- put Mesures, Facinfo, and Patinfo together;
32 proc sql;
33 create view M5_patlist as
34 select a.*b.*c.*
35 from QDFC.M5_patient_list (keep=patid facid &vars year month quarter) a
36 left join saflib.patients (keep=patid surname first_name m_initial ssn) b
37 on a.patid=b.patid
38 join faclib.facinfo_lookupdt (keep=facid network provname provcity DFC_report state wh
39 on a.facid=c.facid
40 order by a.facid, b.surname, b.first_name, a.year, a.month;
41 quit;
```

Figure 4.1b Redeveloped Code (part2)

```
42
43 *-- print out pat lists by facility along with some data formatting: fake patient id, formatting
44 data saflib.M5_plist &dateit.&runby;
45 set M5_patlist;
46 by facid;
47
48 patient_rec_id=_n_;
49 _fname=trim(first_name)||' '||trim(m_initial);
50 _facility=trim(provname)||', '||trim(provcity)||'. '||state;
51 if month<10 then _report_period=strip(year)||' '||strip(month)||' '||strip(quarter);
52 else _report_period=strip(year)||' '||strip(month)||' '||strip(quarter);
53
54 * -- in order to put SSN in xxx-xx-xxxx format;
55 if prxmatch("/\d{9}") then _ssn1=ssn+0; else _ssn1=-;
56 format _ssn1 ssn11.;
57
58 _fn = "&outpath\M5_PatList." || TRIM(facid) || ".txt";
59 FILE writeout FILEVAR=_fn HEADER=newpage LINESLEFT=_remain LINESIZE=80 NOTITLES NOFOOTNOTES;
60
61 * --if new page then put header lines---;
62 if _ramian<11 then put _page_;
63 put @3 Patient_rec_id @15 Surname @49 _fname @64 _ssn1 @80 _report_period @101 firsts @118
64 if last.facid then put "<---END of FILE--->";
65
66 drop _; network state provcity dfc_report provname;
67 return;
68 * --if new page then put header lines---;
69 newpage;
70 put "CONFIDENTIAL: Patients included in the Adult Uncorrected serum calcium > 10.2 mg/dL meas
71 put "Quarterly Dialysis Compare-Preview for July, 2016 report.";
72 put "MMM Certification Number=" facid "Facility=" _facility;
73 put;
74 put @127 "Patient";
75 put @69 "Social" @105 "First" @130 "Age" @157 "Meets" @172 "Meets" @184 "Eligible";
76 put @1 "Patient ID" @67 "Security" @81 "Report" @88 "Period" @103 "service" @114 "dialysis_"
77 put @3 "Number" @15 "Last Name" @49 "First Name" @69 "Number" @79 "Year Month Quarter" @
78 put;
79 return;
80 run;
81
82
83 ENDSAS;;;;;
```



The key SAS features used in the new application is FILE statement and its option FILEVAR=.

#### FILEVAR=variable

defines a variable whose change in value causes the FILE statement to close the current output file and open a new one the next time the FILE statement executes. The next PUT statement that executes writes to the new file that is specified as the value of the FILEVAR= variable.

<b>Restriction:</b>	The value of a FILEVAR= variable is expressed as a character string that contains a physical filename.
<b>Interaction:</b>	When you use the FILEVAR= option, the file-specification is just a placeholder, not an actual filename or a fileref that has been previously assigned to a file. SAS uses this placeholder for reporting processing information to the SAS log. It must conform to the same rules as a fileref.

The new SAS application produces the same results much more efficiently. Moreover, the log file (Figure 4.2) is nice and clean. It lists all the output files orderly. The runtime is 2.25 minutes (Figure 4.3). Can you believe it? The new application reduced the process time from about 10 hours to about 2 minutes. Comparing to the original version, it saved 99.6% of the runtime.

Figure 4.2 New Log Snapshot One

```
I:\MWSUG16_BB18\M5_DFC_Patient_Lists_quinn_MWSUG.log
10707 NOTE: The file WRITEOUT is:
10708     Filename=\\DISK\quinn_test_output\MBD_PatList_111111.txt,
10709     RECFM=U,LRECL=256,File Size (bytes)=0,
10710     Last Modified=28Apr2016:20:52:01,
10711     Create Time=28Apr2016:16:29:03
10712
10713 NOTE: The file WRITEOUT is:
10714     Filename=\\DISK\quinn_test_output\MBD_PatList_222222.txt,
10715     RECFM=U,LRECL=256,File Size (bytes)=0,
10716     Last Modified=28Apr2016:20:52:01,
10717     Create Time=28Apr2016:16:29:03
10718
```

Figure 4.3 New Log Snapshot Two

```
I:\MWSUG16_BB18\M5_DFC_Patient_Lists_quinn_MWSUG.log
59500 NOTE: The data set SAFLIB.M5_PLIST_201607_QUINN has 6554484 observations and 19 variables.
59501 NOTE: Compressing data set SAFRECC.M5_PLIST_201607_QUINN decreased size by 42.20 percent.
59502     Compressed is 86102 pages; un-compressed would require 148966 pages.
59503 NOTE: DATA statement used (Total process time):
59504     real time          2:02.22
59505     cpu time           1:45.83
59506
59507
59508 196
59509 197
59510 198           ENDSAS;;;;;
59511
59512 NOTE: SAS Institute Inc., SAS Campus Drive, Cary, NC USA 27513-2414
59513 NOTE: The SAS System used:
59514     real time          2:24.67
59515     cpu time           1:46.53
59516
```

## V. CONCLUSIONS

This real case study shows us that programming with efficiency in mind can make a great difference:

- 79 lines vs. 150 lines
- 1 step vs. 6,384 steps
- 22,518,989 vs. 61,852,446 records processed
- 00:02:30 vs. 11:02:17 (hh:mm:ss). Process time saved 99.62%.

Besides **raising awareness** for programming efficiency and introducing a **log analysis utility**, this case study presented two important suggestions to promote the performance of SAS applications.

First, developing a better SAS application requires a better understanding of the problem the application is to solve; once the problem is well understood, the programmer's problem solving skills help to design the right algorithm to tackle the problem. This design phase should involve as many knowledge and skills as possible, such as analytics, modular and parallel, data structure, logic/abstract/model and system thinking, etc.

Second, the application developer/programmer's SAS knowledge, experience, and skills also play an important role in programming efficiency. Here are some general SAS programming tips that can be usefully to improve application performance: use as fewer steps as possible if applicable; combine steps/remove unnecessary steps; process only the required variables and observations; avoid complex macro if you can; use simple/non-macro coding effective techniques; do not fall in love with your "hammer", know and pick the right tool to use; be machine, human and computing environment friendly.

## REFERENCES

SAS Online Documentations for SAS 9.2, 9.3 and 9.4. (<http://support.sas.com/documentation>)

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## CONTACT INFORMATION

Your comments and questions are valued and encouraged. Contact the author at:

Lingqun Liu  
University of Michigan  
Kidney Epidemiology and Cost Center  
1415 Washington Heights, Suite 3645 SPH I  
Ann Arbor, MI 48109-2029  
Email: [lqliu@umich.edu](mailto:lqliu@umich.edu)  
<http://www.kecc.sph.umich.edu/>

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