SAS Programming Efficiency: Tips, Examples, and PROC GINSIDE Optimization

Lingqun Liu, University of Michigan
MISUG, Feb 2018

-

Outline

- This paper first explores the concepts of efficiency.
- Then reviews some relevant materials and tips available online.
- Examples of efficient programming.
- PROC GINSIDE optimization.

Efficiency: Concepts

- Computer resources
- Human resources
- SAS OPTIONS: STIMER, FULLSTIMER
- Time, time, time:
 - Less I/O time,
 - Less CPU time,
 - Less human time
- Two principals/strategies: Do the right thing, and do it right.

3

Programming Efficiency Tips

- Google search for "SAS efficiency"
- Presentations at MSUG meetings

"Leave Your Bad Code Behind: 50 Ways to Make Your SAS Code Execute More Efficiently", by William Benjamin, June 2017 one-day conference.

"SAS Advanced Programming with Efficiency in Mind: A Real Case Study", by Lingqun Liu, Feb 2017 meeting.

"Quick Hits - My Favorite SAS Tricks", by Marje Fecht, May 2013 one-day conference.

"Utilizing SAS for Efficient Coding", by Michelle Gayari, November 2009 meeting.

Peri How to Write Efficient SAS Programs in Eleven Easy Steps!

publicheathrsciences wustliedu/en. /-/- //ASQTreangra. Optimize-Your-Gode ashx =
MAKE THENDS CALEER FOR VOURSELE FETICHESEY ALSO MAKEN WORKING SMARTER - te-"GREED

- save-code and reuse in later! - Collaborate with your co-workers to share lips and suggestions. All

engoluthy to share ideas. Some weep SAS Gode forester esusability: Maken Estary. Stored processes

dasug dartmouth edu/wp-content/uploads/OptimizeSASCode_Long.pdf = 1
10.MARE THINGS EASIER FOR YOURSELF EFFICIENCY ALSO MEANS WORKING SMARTER! - Be
10.MARE THINGS EASIER FOR YOURSELF SHOP CONTROL OF THE STATE OF

SAS Programming Efficiencies https://www.ssc.wisc.edu/sscc/pubs/4-3.pdf +

SAS Programming Efficiencies. Last revised: 04-21-99. The purpose of this document is to provide tips for improving the efficiency of your SAS programs. It suggests coding techniques, provides guidelines for their use, and compares examples of acceptable and improved ways to accomplish the same task. Most of the tips.

SAS speed tips - Data Savant Consulting www.datasavantconsulting.com/roland/speedtips.html Feb 2, 2016 - if you transitioned over from another efficient to the consulting time of t

Feb 22, 2016 - If you transitioned over from another efficient batch language such as COBOL then seeking data handling efficiency for SA3 would be natural and the first thing you would apply yourself is because you came from a background with eimilar constraints. But, If your programming background was with ...

Terrificient Techniques and Tips in Handling Large Datasets - Lex Jansen www.legiansen.com/wass/2011/coders/Papers, Kuang_J_75005.pdf *
Efficient Techniques and Tips in Handling Large Datasets. Silving Kuang, Kelly Blue Book Inc., Irvine, CA. ABSTRACT, When we work on millions of records, with hundreds of variables, it is crucial how we are processing our data. To make SAS. 6 really ROCK, we need to pay more attention to SAS. 6 program efficiency ...

| Introduction to efficiency techniques in SAS programmation - PhUSE www phuse earlydemional asprofyper-emaßdocid- 1859 + Introduction to efficiency techniques in SAS programmation. Van Holle Uncel ... Efficiency techniques in SAS programmation. Van Holle Uncel ... Efficiency techniques are more and more crucial for several reasons: - Data grow larger in every pharmaceutical field: clinic trials given project, ... Bolutions: - Practical tips to customize a SAS session. - Piffrorming delayed.

SAS Programming Tips: A Guide to Efficient SAS Processing

.

Programming Efficiency Tips

if vars[i]>1 then do;

A= var[i]*amount*cmmisions;

New Tips Use array to reduce coding ... set sample; array vars {3} vara varb varc; do i=1 to 3;

Original tips #16-#17 presented at MISUG June 2017 one-day conference

#	This works:	This is uses less code:
16	Computers are good at doing the same thing over and over again. But, programmers do not like to code the same thing over and over again. Take this	One way to rewrite the code on the left is to use subroutines. The following is one example.
	code for an example.	Data test;
	Data test;	Set sample;
	Set sample;	Temp =vara; If vara >1 then link test:
	If vara > 1 then do:	Troid Troid min tool,
	A = vara * amount * commission:	Temp =varb:
	B = A / 2:	If varb >1 then link test:
	C = A + B;	
	End;	Temp =varc;
	If varb > 1 then do;	If varc >1 then link test;
	A = varb * amount * commission;	
	B = A / 2;	Return; * end of the main datastep;
	C = A + B;	
	End; If varc > 1 then do:	* this subroutine code is executed three times;
	If varc > 1 then do; A = varc * amount * commission:	Test:
	B = A / 2:	A = Temp * amount * commission;
	C = A + B:	B = A / 2:
	End:	C = A + B:
		Return: * end of the subroutine:
	Run;	Run;
17	This is the same code as the previous example, but	The use of a macro to do the same task over and over
	a second solution is presented that makes the code	again works well too

3) Use temp variable to reduce coding

2) Use IF ELSE to avoid wasting extra CPU time

...
set sample;
if varc > 1 then _temp = varc;
else if varbo! then _temp = varb;
else if varao! then _temp=vara;
if _temp>1 then do;

4) Use temporary variable and IFN() function to reduce coding

temp=ifn(varc>1,varc,ifn(varb>1,varb,ifn(vara>1,vara,.))); | a second solution is presented that makes the code | again works well too.

Programming Efficiency Tips

New tips

```
1) Use built-in function REPEAT() to simplify the code.
```

1) Use array to simplify the code

end;

ray (0 simplify the code array conds {31} cond_1 - cond_31; do i=31 to 1 by -1; if conds[i] = i then substr (flag, 32-i,1) ='1';

1) Use LENGTH statement and CATS() to simplify the code

do i=1 to 31; flag=cats(conds[i]=i,flag);

What if the conditions ${\tt conds[i]=i}$ are changed and there is no pattern? A TEMPORARY array will do the trick.

An array will 00 the UTCK.

array _cs {31} _temporary_ (1 2 3 ... 4 5 6);

do i=1 to 31;

flag=cats(conds[i]=_cs[i],flag);

end:

Original tip #48 presented at MISUG June 2017 one-day conference

```
Set my_sas_file;
   If (cond_1 = 1) then substr(flag,31,1) = '1';
If (cond_2 = 2) then substr(flag,30,1) = '1';
If (cond_3 = 3) then substr(flag,29,1) = '1';
   *. . . more conditions . . . ;
   If (cond_29 = 4) then substr(flag,3,1) = '1'; If (cond_30 = 5) then substr(flag,2,1) = '1'; If (cond_31 = 6) then substr(flag,1,1) = '1';
   Conditions_flag = input{flag,ib4.}; * convert the flags to a real numeric variable; * IB4. informat is Integer Binary for 4 bytes.; * It knows there were 31 binary lights not 32;
run;
```

- How to check missing values of all variables in a data set
- How to identify the new or changed records
- How to identify common variables in multiple data sets
- Use the right SAS built-in functions

7

Programming Efficiency Examples

• Check missing values of all variables in a data set

```
ods output OneWayFreqs = _checking_missing_
proc format;
                                        (keep=table frequency percent cumfreq: cumpercent f_:);
         value $miss
                                    proc freq data= _test_;
         '',' '='c_missing'
                                       table _all_/missing;
                                        format _numeric_ miss. _CHARACTER_  $miss.;
         other='c_non-missing'
                                   run;
                                    ods output close;
         value miss
         .='n_missing'
                                    data _missing_;
         other='n_non-missing'
                                        length table $32;
                                        set _checking_missing_ ;
run;
                                        length formated $30;
                                        formated = cats(of f_:);
                                        table=substr(table,7);
                                        drop f_:;
                                    run;
```

• Check all variables in a data set w/o using %macro loop

Similarly, this technique can be used to summarize all numeric variables in a data set.

```
proc means data=_test_
noprint;
    var _numeric_;
    output
out=_all_mean_ n=
nmiss= mean=/autoname;
run;
```

Create Frequency table for all variables in a data set.

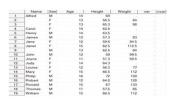
```
ods output OneWayFreqs = _freq_all_ (keep=table
frequency percent cumfreq: cumperc: F_:);
proc freq data=sashelp.class;
   table _all_/missing;
run;
ods output close;

data freq_all (keep = varname value freq: cum:);
   set _freq_all_ (rename = (table=varname));
   value=cats( of F_:);
   varname=substr(varname,7);
run;
```

9

Programming Efficiency Examples

• Check missing values of all variables in a data set



	varname	N	NMiss	Mean
1	Age	19	0	13.315789474
2	Height	19	0	62.336842105
3	Weight	15	4	99.533333333
4	var	0	19	

		Table	Freq	Percent	CumFreq	CumPerc	F_Name	F_Sex	F_Age	F_Height	F_Weight	F_var	F_cvar
	1	Table Name	3	15.79	. 3	15.79	c_missing						
	2	Table Name	16	84.21	19	100.00	c_non-missing						
2	3	Table Sex	19	100.00	19	100.00		c_non-missing					
-	4	Table Age	19	100.00	19	100.00			n_non-missin				
	5	Table Height	19	100.00	19	100.00				n_non-missing			
	6	Table Weight	4	21.05	4	21.05					n_missing		
	7	Table Weight	15	78.95	19	100.00					n_non-missing		
	8	Table var	19	100.00	19	100.00						n_missing	
	9	Table cvar	19	100.00	19	100.00							c_missing

	table	Frequency	Percent	CumFrequency	CumPercent	formated
1	Name	3	15.79	3	15.79	c_missing
2	Name	16	84.21	19	100.00	c_non-missing
3	Sex	19	100.00	19	100.00	c_non-missing
4	Age	19	100.00	19	100.00	n_non-missing
5	Height	19	100.00	19	100.00	n_non-missing
6	Weight	4	21.05	4	21.05	n_missing
7	Weight	15	78.95	19	100.00	n_non-missing
8	var	19	100.00	19	100.00	n_missing
9	cvar	19	100.00	19	100.00	c_missing

• Identify the new or changed records

Use DATA-MERGE

Use SQL-set operation

11

Programming Efficiency Examples

• Identify common variables in multiple data sets

```
proc sql;
  create table _common as
  select * from a where 0
  union corr
  select * from b where 0
  union corr
  select * from c where 0
;
quit;
```

SQL set operation overlays columns that have the same name in the tables, when used with EXCEPT, INTERSECT, and UNION, CORR (CORRESPONDING) suppresses columns that are not in all of the tables.

• Use the right SAS built-in functions

```
Old:
    text = TRANWRD(TRANWRD(TRANWRD(TRANWRD(htmltext,'>','>'),'<','&lt;'),
    '&','&amp;'), '"', '&quot;'), "'", '&apos;');

New:
    text = HTMLDECODE(htmltext);

Old:
    initial = substr(first_name,1,1)||substr(last_name,1,1);

New:
    initial = first(first_name)||first(last_name);

Old:
    cdate = put(year(datepart(datetime())), f4.) || put(month(datepart(datetime())), z2.);

New:
    cdate = put(today(),yymmn.);</pre>
```

13

Optimize PROC GINSIDE

- PROC GINSIDE overview
- An application: find Blocks for Zip code centers
- PROC GINSIDE performance
 - Large data sets
 - Intensive computations
- Optimize PROC GINSIDE
 - Reduce map data sizes SELECT statement
 - Preliminary search Block limits of XY coordinates
 - Search within the selected Blocks only %macro Loop to create ZIP specific map data set and run PROC GINSIDE for each ZIP.

PROC GINSIDE overview

PROC GINSIDE was first introduced in SAS 9.2. "The new GINSIDE procedure determines which polygon in a map data set contains the X and Y coordinates in your input data set. For example, if your input data set contains coordinates within Canada, you can use the GINSIDE procedure to identify the province for each data point."

PROC GINSIDE is a application of the **point-in-polygon** (**PIP**) problem.

- An application: find Blocks for Zip code centers
 - CENSUS Block

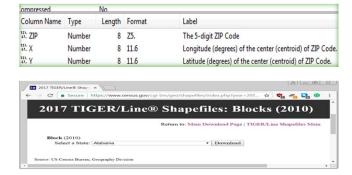
https://www.census.gov/newsroom/blogs/random-samplings/2011/07/what-are-census-blocks.html



15

Optimize PROC GINSIDE

- A application: find Block for Zip code centers
 - ZIP code and CENSUS Block data sets



- Texas has about 2600 ZIP codes and 914,231 Census Blocks
- 1. Convert Shapefile to SAS MAP data

MAP.MAP_48_BLOCK has 43,353,186 observations and 4 variables.

2. ZIP code data

```
data zip_48;
    set sashelp.zipcode(keep= x y zip state);
    where state=48;
run;
```

17

Optimize PROC GINSIDE

• Performance w/o optimization

• New algorithm step 1

```
* Algorithm step 1: use limits of each geoid10 to match zip codes;
proc sql;
    create table block_&fp._limit
as select geoid10 ,max(y) as max y ,min(y) as min y ,max(x) as max, y ,max(x) as max, ,min(x) as min x from map.block afp group by geoid10; quit;
proc sql;
   create table gzinside_1_&fp
      select a.*
     select a.*
,b.geoid10
,count(distinct geoid) as ct_match
from zip.zipcode (keep=zip x y state where=(state=&fp)) a
left join block_&fp. limit b
on a.y between b.min_y and b.max_y
and a.x between b.min_x and b.max_x
group by zip; quit;
```

ct_match	Frequency	Percent	Cumulative Frequency	Cumulative Percent
0	3	0.12	3	0.12
1	952	36.62	955	36.73
2	958	36.85	1913	73.58
3	417	16.04	2330	89.62
4	166	6.38	2496	96.00
5	63	2.42	2559	98.42
6	23	0.88	2582	99.31
7	11	0.42	2593	99.73
8	6	0.23	2599	99.96
9	1	0.04	2600	100.00

Some states, like OK, can have more than 55% matched with ct_match=1.

19

Optimize PROC GINSIDE

New algorithm step 2

```
* Algorithm step 2: use proc ginside for zip codes with multiple matched;
%macro ginside(zip);
   create table map &zip
from map.map_&fp._block a, map.gzinside_1 b where b.zip=&zip and a.geoid10=b.geoid10; quit;
proc ginside
    data= map.zip_sfp (where=(zip=szip))
    map = map_szip
    out = ginside_szip;
    id geoidl0;
*mend ginside;
Each ZIP/GINSIDE took only 0.02 ~ 0.04 seconds.
```

run; Cumulative Cumulative Frequency Percent 36.62 63.27 99.88

data null_;
 set map.gzinside_1;
 where ot_match>1;
 call execute(catt('\$ginside(', zip,');'));

ta map.gzinside_: **p*,
set ginside_:
 map.gzinside_1 (in=a where=(ct_match<2));
 if a then rc=1; else rc=2;
 if a and geoid10='' then rc=9;</pre>

data map.gzinside &fp;

proc freq data=map.gzinside_&fp;

New algorithm step 2:

• %macro Loop: create and process ZIP specific data
Applications that break and process data sets chunk by chunk are not efficient if the data
sets can be processed as a whole, because it increases I/O operations. An example can be
found in the paper presented at MISUG Feb 2017 meeting. Here the situation is different.
%macro loop is efficient.

Data-oriented

Instead of searching among about 914,231 polygons in Texas Block data set (43,353,186 observations), the new algorithm search only among 2 to 9 polygons for each zip code. It runs much faster since it reduces lots of CPU time and I/O time.

21

Optimize PROC GINSIDE

• Results and improvement

	records	runtime - PC	runtime - Linux
w/o optimization	310 ~ 780 ZIP codes in Texas		31~33 hours
	2600 ZIP in Texas	> 3 days, job killed	<30 minutes
	4356 ZIP codes		135 hours
optimization	2600 ZIP codes	< 6 minutes	3 minutes
	41k ZIP codes	< 8 hours, 1 hour w	< 30 minutes
		reuse of the limits files	

Summary of the optimization

- 1. Use the **select** statement to reduce map data file size.
- 2. Use Block **limit** data sets (that have way much less observations than the original Block data sets) to perform first match.
- 3. Use **ZIP specific map data** sets to enormously reduce the search range of GINSIDE procedure. Instead of searching within 914,231 blocks, GINSIDE only searches within about 5000 blocks overall.
- 4. In short, it reduces a large number of the processed records; therefore, it reduces I/O and CPU time. The improvement is significant.

23

Another Optimization Example

- 1. Medicare Part D claim data and patient data
- 2. Code is shorter, easier to understand (user friendly) Less than 80 lines. (original one has 185 lines)
- 3. Run faster

One-drug job run time less than 1 hour (original one took 13^3 + hour) Three-drug job run time less than 3 hours (original one took 72^9 6 hours)

4. Algorithm: simplified, all in one batch/bunch process.

Avoid %macro loop. Only 5 steps (original one has around 1000 steps)





Questions and Comments

THANK YOU!

Contact: Iqliu@umich.edu