Paper CC07 SQL or not SQL When Performing Many-to-Many Merge - A Realistic Example

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ABSTRACT

PROC SQL is a powerful tool for SAS programmers. Choosing between PROC SQL and a DATA step is a matter of preference in most cases, but in some cases, PROC SQL is a better choice in terms of performance and efficiency. This paper will illustrate the differences between PROC SQL and DATA step when performing a many-to-many merge, and will elaborate on situations when PROC SQL is a preferred choice. This paper is targeted at an intermediate audience.

SAS Version 8.2 or 9.1, Windows XP or UNIX, Intermediate Level

KEYWORDS

PROC SQL, MANY-TO-MANY MERGE, DATA STEP, SQL JOIN, MERGE.

INTRODUCTION

SQL (Structured Query Language) is a data manipulation language common to relational Databases, such as Oracle, SQL Server, etc. PROC SQL is SAS's Implementation of SQL. PROC SQL is a powerful SAS procedure that combines the functionality of the SAS DATA step with SQL Language. PROC SQL can sort, subset, merge and summarize data – all at once. PROC SQL can tackle most of the tasks that a DATA step can perform. In this paper, the differences between PROC SQL and the DATA step will be briefly explained for a many-to-many merge. The focus is on elaborating realistic situations when PROC SQL is a preferred choice between these two methods.

PROC SQL vs. DATA STEP

When we want to combine data horizontally, PROC SQL joins and DATA step match-merge can produce identical results when performing a one-to-one merge or a one-to-many merge. But when we do a many-to-many merge, PROC SQL and DATA step merge will produce different results. Let's see a simple example:

```
data A;
      x=1; y=10; output;
      x=1; y=15; output;
      x=1; y=18; output;
      x=3; y=20; output;
run;
data B;
      x=1; z=15; output;
      x=1; z=12; output;
      x=3; z=30; output;
run;
proc sort data=a; by x; run;
proc sort data=b; by x; run;
data c1;
      merge a(in=aa) b(in=bb);
      by x;
      if aa;
run;
```

```
NOTE: MERGE statement has more than one data set with repeats of BY values.
NOTE: There were 4 observations read from the data set WORK.A.
NOTE: There were 3 observations read from the data set WORK.B.
```

NOTE: The data set WORK.Cl has 4 observations and 3 variables.

```
NOTE: Table WORK.C2 created, with 7 rows and 3 columns.
```

The following are datasets C1 and C2; they are quite different:

Dataset C2:	Dataset C1:
X Y Z	$\mathbf{X} \mathbf{Y} \mathbf{Z}$
1 10 15	1 10 15
1 15 15	1 15 12
1 18 15	1 18 12
1 10 12	3 20 30
1 15 12	
1 18 12	
3 20 30	

From the above example, we can clearly see that when there is a many-to-many match on the values of the BY variable, a DATA step match-merge probably does not produce the desired output because the output data set will not contain all of the possible combinations of matching observations. On the other hand, when the datasets are joined using PROC SQL, each match appears as a separate observation in the output dataset, this probably is a more desired output.

Clinical Trial Example

Let's illustrate this with a real life example. Suppose we need to create creatinine clearance records for each visit. In the data, there is a measurement of serum creatinine by using the Cockroft-Gault formula:

```
Males: CCr (ml/min) = \frac{(140\text{-age}) \text{ x body weight (kg)}}{\text{creatinine (umol/L) x 0.814}}
Females: CCr (ml/min) = \frac{(140\text{-age}) \text{ x body weight (kg) x 0.85}}{\text{creatinine (umol/L) x 0.814}}
```

In order to calculate the creatinine clearance, a patient's weight is needed. Weight measurements are stored in the VITAL dataset, and there are multiple weight measurements for different dates. We need to merge the LAB dataset with the VITAL dataset to get the last available weight on or before serum creatinine measurement date. This is a many-to-many merge, PROC SQL can finish this job quite easily while simply using DATA step match merge will not produce the desired result.

The following code will produce the LAB and VITAL datasets. First, a fake lab serum creatinine dataset is created.

```
data lab;
  length lbtestcd $ 10 lbstresu $ 20 ;
  retain lbtestcd "CREAT" lbstresu "UMOL/L";
  do usubjid=1001 to 6000;
    if ranuni(0)<0.5 then sexcd=1;
    else sexcd=2;
        age=floor(47+sqrt(81)*rannor(1688));
        lbdt=floor(14535+sqrt(219950)*rannor(1688));
        do visitnum= 1 to 6 ;
        lbdt=lbdt+30;</pre>
```

```
format lbdt yymmdd10.;
               label
                 usubjid = 'Unique Subject Identifier'
                           ='Sex Code 1=male 2=female'
                 sexcd
                           ='Age in Years'
                 age
                 lbdt
                           ='Parameter Measurement Date'
                 lbstresn ='Parameter Value'
                 lbtestcd ='Parameter Name'
                 lbstresu ='Parameter Unit'
                 visitnum ='Visit identifier (numeric)'
                 output;
           end;
       end;
run;
NOTE:
      The data set WORK.LAB has 30000 observations and 8 variables.
proc sort data=lab;
       by usubjid lbdt;
run;
In order to create another vital sign dataset with reasonable date values, we need to get the first lab date for each
patient.
proc sort data=lab;
       by usubjid lbdt;
run;
data firstlab;
       set lab;
       by usubjid lbdt;
       if first.usubjid;
       keep usubjid lbdt;
run;
Now we are ready to create a fake vital sign dataset.
data vital(drop=i);
       length vstestcd $ 10 vsstresu $ 20;
       retain vstestcd "WEIGHT" vsstresu "KG";
       do usubjid=1001 to 6000;
           do i = 1 to 10;
                      vsstresn = floor(74.5 + sqrt(276) * rannor(1688));
              label
                   vstestcd = 'Vital Parameter Name'
                   vsstresn ='Vital Parameter Value'
                   vsstresu ='Vital Parameter Unit'
                      output;
           end;
       end;
run;
proc sort data=vital;
       by usubjid;
run;
data vital(drop=lbdt);
       retain vsdt;
```

merge vital(in=a) firstlab(in=b);

lbstresn = 70+sqrt(200)*rannor(1688);

```
by usubjid;
   if a;
   if first.usubjid then vsdt=lbdt-1;
   vsdt=vsdt-10;
   format vsdt yymmdd10.;
   label vsdt ='Vital Parameter Measurement Date';
run;
```

```
NOTE: The data set WORK.VITAL has 50000 observations and 5 variables.
```

Now that we have two fake datasets ready, we can proceed to merge them in order to calculate the creatinine clearance for each visit, which has a non-missing serum creatinine record and has an available weight on or before such visit. The following PROC SQL code performs the many-to-many merge:

```
proc sql;
   create table crcl (label='Creatinine Clearance Dataset' )as
   select
       "CRCL" format $10. as lbtestcd label='Parameter Name',
       "ML/MIN" format $20.as lbstresu label='Parameter Unit',
       a.usubjid,
        a.sexcd,
        a.age,
        a.lbdt,
        a.visitnum,
                                        label='Parameter Measurement',
       a.lbstresn as creat
       b.vsstresn as weight
                                        label='Weight (KG)',
                                        label='Weight Measurement Date'
       b.vsdt
    from lab as a left join vital as b
       on a.usubjid=b.usubjid and a.lbdt>=b.vsdt
    where a.lbstresn ne .
    order by lbtestcd, lbstresu, usubjid, visitnum, lbdt, vsdt
quit;
```

As illustrated in the above simple example, the PROC SQL join contains all possible combinations of matching observations. .

Here is one way to accomplish the same task by using a DATA step:

```
data crcl_;
    set lab(keep=usubjid visitnum lbdt lbstresn sexcd age
            rename=(lbstresn=creat)) end=eof;
    flag=0;
    do pi=1 to numobs;
        set vital(rename=(usubjid=subid vsstresn=weight))
                    point=pi nobs=numobs;
        if usubjid=subid and lbdt >= vsdt then do;
            flag=1;
            output;
        end;
        else if pi=numobs and flag=0 then do;
            weight=.;
            output;
        end;
    end;
    if eof then stop;
    drop subid flag;
```

```
NOTE: The data set WORK.CRCL_ has 300000 observations and 10 variables.

NOTE: DATA statement used:
real time 12:51.06
cpu time 12:48.51
```

Comparing the performance of PROC SQL with a DATA step method, there is a huge CPU time difference for 5000 patients, which is typical population size for an ISS study, PROC SQL is much more efficient in such case.

We then need to sort them to get the last available weight, and then calculate the creatinine clearance value.

```
proc sort data=crcl_;
      by usubjid visitnum lbdt vsdt;
run;
data crcl_(drop=creat weight);
      set crcl_;
      by usubjid visitnum lbdt vsdt;
      if last.visitnum;
      lbstresn=0.85**(sexcd-1)*(140 - age)*weight/(creat*0.814);
      if lbstresn ne .;
run;
data crcl(drop=creat weight);
      set crcl;
      by lbtestcd lbstresu usubjid visitnum lbdt vsdt;
      if last.visitnum;
      lbstresn=0.85**(sex-1)*(140 - age)*weight/(creat*0.814);
      if lbstresn ne .;
run;
```

```
NOTE: There were 300000 observations read from the data set WORK.CRCL.
NOTE: The data set WORK.CRCL has 30000 observations and 8 variables.
```

The newly created creatinine clearance dataset is now ready to be concatenated together with the original lab data to create an updated lab dataset ready for statistical analysis.

```
data lab;
     set lab crcl;
run;
```

```
NOTE: There were 30000 observations read from the data set WORK.LAB.

NOTE: There were 30000 observations read from the data set WORK.CRCL.

NOTE: The data set WORK.LAB has 60000 observations and 8 variables..
```

The above example shows that PROC SQL merge works well under certain circumstances, which requires a many-to-many merge. In such cases, simply using DATA step match-merge will not get the job done, but would require additional DATA step manipulations, which may involves either lengthy coding or short but resource intensive coding.

CONCLUSION

Combining data horizontally is very common in the SAS programming world. For a one-to-one or one-to-many merging, choosing PROC SQL or a DATA step match-merge is just a personal preference, since these two methods will give you identical results. However, when performing a many-to-many merge, PROC SQL is the preferred choice.

REFERNCE

Cockroft-Gault formula to calculate creatinine clearance: http://en.wikipedia.org/wiki/Renal_function

 $Hu, Weiming.\ 2004.\ ``Top\ Ten\ Reasons\ to\ Use\ PROC\ SQL".\ Proceedings\ of\ the\ Twenty-Ninth\ Annual\ SAS\ Users\ Group\ International\ Conference,\ paper\ 42$

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