# Putting the Meta into the Data: Managing Data Processing for a Large Scale CDC Surveillance Project with SAS®

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#### **ABSTRACT**

There are myriad epidemiological and surveillance studies ongoing due to the pervasive COVID-19 pandemic, often embodying the definition of "big data" with thousands of participants, variables, and lab samples. Data can be derived from many different streams in a given study, for example: REDCap software, electronic medical records (EMR), chart abstraction, laboratory records, etc. Different contractors can be managing distinct aspects of the same project, the data is changing minute to minute, and the deliveries are required at a fast and furious pace. Wrangling all the different data sources requires robust data management routines, and SAS® can help, with tools to obtain data via APIs and PROC HTTP, metadata resources, and programming techniques. This paper and presentation will outline best practices for managing multiple aspects of large scale CDC surveillance projects, using SAS.

# **INTRODUCTION**

Any big study involves data acquisition, deliverable schedule, data and programming management concerns, ever-changing data structures, the dangers of programming "silos", data and code quality concerns, tracking and communication, and most of all time concerns. This paper will look at two large scale, multi-site CDC surveillance studies involving COVID and flu testing, and other health metrics. Both projects are with the CDC, involve distinct aspects of COVID 19, and the CDC is heavily involved in analytic planning and specifications. The EMR project had monthly deliveries, while the REDCAP project has weekly to biweekly deliveries, and in rare cases more than one delivery a week. Results from the REDCAP project have gone directly to CDC and the White House, so the stakes are extremely high in terms of quality assurance and rapid turnaround. The EMR project was designed to assess the effect of COVID 19 on pregnant women and children and had three sites across the United States. The REDCap project (one of a number of different projects in house) was designed to monitor the ongoing effects of the COVID 19 pandemic on first responders in the US, and has numerous sites and REDCaps.

If you are not familiar with REDCap, it is a browser-based, metadata-driven EDC software and workflow methodology for designing clinical and translational research databases. Many institutions use REDCap to develop online surveys and databases. There are millions of projects using this resource. It definitely has a pretty clunky interface with regards to SAS, a little better with R, but we endeavor to make it work.

#### **SCENARIOS**

We will discuss how we use SAS to help wrangle these COVID related data bases. For scenario 1, we will dive into data dictionary read in and uses in labeling, formatting, etc. For scenario 2, we will look at comparisons of data dictionaries and formats.

As noted above, these projects have different study designs: Scenario 1 has a combination of REDCap survey data and EMR data, while Scenario 2 has a combination of REDCap survey data and serology test results. These two projects have common and differing data management and SAS programming challenges. Commonalities for both of these scenarios are: a vast amount of data from different data sources, multiple data areas, and multiple analytic agendas; quick turnaround deliveries (monthly on one project, weekly+ on another project); and very high visibility (results going into White House briefs) which necessitates actionable quality assurance.

A cornerstone for big projects like these is data management. My company mandates data management and code quality assurance plans for every project. Key data management tools that we use on a daily basis include: data receipt and delivery logs; shared process trackers; details of the data management and code quality plans, staffing plans, and last, but not least, data dictionaries. We will drill down on the use of data dictionaries to drive much of our processing for both projects, creating files and processes

from the metadata contained in data dictionaries and files to drive the provision of variable and value labels, order files, assign variable prefixes and split out subsets of files, as well as drive the creation of format assignment statements and reporting for file subsets.

#### **SCENARIO 1**

A data dictionary for a file based on Electronic Medical Records (EMR) contains variables which represent an unknown number of COVID-19 tests for an unknown number of infants – there is no way to know in advance how many iterations of the COVID test variable will exist in the actual data file from medical entities. In addition, variables in this file may exist for three distinct groups (pregnant women, postpartum women, and infants), with PR, PP and IN prefixes, respectively. We will demonstrate how to ingest data dictionaries to collect metadata for this project allowing identification of groups for processing, and drive label (and value label) description creation for iterated (and other) labels using SAS functions, PROC FORMAT, AND DATA step processing, as well as other utilities.

Data dictionaries that are constructed via a collaboration between the CDC, study sites, and project staff are provided to study sites, who return extracted electronic medical record (EMR) data to us for processing.

There are two separate data dictionaries used for this project, for person-level and visit-level variables. The data dictionaries in use for the study have multiple tables and thousands of variables. Individual variables have up to two # signs (iteration flags), appear in a single row in the data dictionaries, and may occur more than sixty times. Variables to the right of the screenshot shown below indicate whether the variable is present for pregnant women, postpartum women, and/or infants. There are six data files delivered from each site – 2 data dictionaries x 3 populations. We need variable and value labels for each of the six data files, so we need to build LABEL statements, a format library, and format assignments. We receive data on an ongoing basis (monthly), so the solution needs to be as efficient and data driven as possible. Iterated variables, which appear with pound signs in the data dictionaries, appear with numbers in the actual data and we need a programmatic answer to match the variables in the data dictionaries to the variables in the data files.

Solutions presented include ingesting data dictionaries to collect metadata allowing identification of groups for processing, and drive processing with information derived from data dictionaries.

# Data dictionary read in

We read in the separate tabs in the workbook in the data dictionary and collect information to be used for variable names, labels, and value label, as well as other information used solely for data processing. Note the red # signs, the variable description, which will be turned into a label, and the variable values, which will be turned into value labels (a format) in Figure 1 shown below.

Challenges for read in include: multiple tabs in each Excel workbook: different starting rows in each tab, requiring reading in specific ranges; clean up of special characters (tabs, carriage returns), disaggregation of some fields, making sure errors encountered in read in are addressed; and the need to collect information on two levels (variable information vs value label information).



# Figure 1. Sample Data Dictionary

Since multiple tabs are read in with the same structure on each tab, we take advantage of macro processing to read in our file metadata. We use ranges in our PROC IMPORT, storing valuable metadata about our incoming data. Note that it is possible to have two level range names, similar to library names, by preceding the range with the tab name and separated by a dot.

First, the LENGTH function is used to calculate the length of "variable". The length of variable names is limited to thirty-two characters and the name an of iterated variable may exceed the limits. The data dictionary is a living document, and if any overlong variables that exist once prefixes and iterated counts are added to their base, the spelling is adjusted. Identification variables are exempt from prefixes. When a file is first processed, variables, with the exception of ID variables, have prefixes added, using the CATS function. Additionally, label strings are created by concatenating prefixes and existing variable descriptions using the CATX function.

```
data labels&tabn.;
    length label labelstr $ 300 variable_type $ 8;
    set &tabnm (keep=variable_: pw_preg pw_pp inf
    where=(variable_name ne '' and variable_description ne ''));
    label=catx(": ","&tabnm.",variable_description);
    labelstr=cats(variable_name,'="',label,'"');

    variable_length=length(variable_name);
    length_flag=(variable_length+7 GT 32);
    label variable_length="Length of Variable"
    length flag="Current Variable_Length + 7 exceeds 32";
```

In preparation for iteration, we use the INDEXC function to find the location (or existence) of # signs. If we wanted to look for a string (as we do later on) we can use the INDEX function. We use COUNTC to count how many times an iteration flag occurs in a variable name. Multiple iterations of a variable can occur if, for example, multiple neonates in a single pregnancy have multiple virus tests.

```
/* find out the # of iterations within a variable name */
iteration_flag=(indexc(variable_name,'#') gt 0);
iteration_count=countc(variable_name,'#');

label iteration_flag="Binary: Variable iterations"
iteration_count="# of iteration points within variable name";
run;

%mend;
%imptabs(tabn=1, tabnm=Identifiers, intab=Identifiers, startrow=4, endcol=H);
```

#### **Iteration**

It is relatively simple to replace a single iterator, in this case, a #, in a variable name. It is more complicated to replace two or more iterators, especially if you do not know how many iterations there are. SAS functions process one variable transformation at a time – that is, they stop after completing a single operation on a string. After confirming the existence of a # sign in a variable and finding its position using the INDEX function, we then use the SUBSTR function to replace the # using a do loop, outputting additional label records for each iteration.

As noted above, we use the COUNTC function to discover how many #s exist in a variable name. You can use functions to discover the number of iterations needed as well in the actual data - including the REVERSE and ANYNUM functions - in the actual data. Additionally, the iteration numbers are added to the label strings using CAT functions. Multiple supplemental label records are created until no more # signs appear in the variable names.

We have thousands of variables, and multiple occurrences of iteration and the need to replace (via the SUBSTR function) items of different lengths. We quickly realized we would need to employ macro loops to manage the different requirements for a number of situations (number of iterations, the "base" of the variables needing to be iterated, one or two iteration symbols, and substring length). Sample code for a simple loop and more complex loop follow below.

# Simple loop

```
%macro do list1(maxiter=1, suffix=neo);
%do i=1 %to &maxiter;
data iter&suffix.1 &i (drop=loc);
    length variable $ 50 labelstr $ 300;
    set formats0 (where=(count(variable, "#")=1 and
index(variable, "IDENTIFIER#") > 0));
    *get the first indexed # location;
    loc=index(variable,"#");
    substr(variable, loc, 1) = "&i";
    labelstr=catt(labelstr," #&i");
run;
%END;
%MEND DO LIST1;
Complex loop
%macro do list2(maxiter=20, suffix=vtst);
```

```
%if &maxiter le 9 %then %do i=1 %to &maxiter;
data iter&suffix.1 &i (drop=loc);
    length variable $ 50 labelstr $ 300;
    set formats0 (where=(count(variable,"#")=1 and
index(variable, "VTST#") > 0));
    *get the first indexed # location;
    loc=index(variable,"#");
    substr(variable, loc, 1) = "&i";
```

```
labelstr=catt(labelstr," #&i");
run;
proc print data=iter&suffix.1 &i (obs=5) noobs;
   var variable labelstr;
%END;
%if &maxiter gt 9 %then %do;
%do i=1 %to 9;
data iter&suffix.1_&i (drop=loc);
    length variable $ 50 labelstr $ 300;
    set formats0 (where=(count(variable,"#")=1 and
index(variable, "VTST#") > 0));
    *get the first indexed # location;
    loc=index(variable,"#");
    substr(variable, loc, 1) = "&i";
    labelstr=catt(labelstr," #&i");
run;
proc print data=iter&suffix.1 &i (obs=5) noobs;
   var variable labelstr;
run;
%END;
%do i=10 %to &maxiter;
data iter&suffix.1 &i (drop=loc);
    length variable $ 50 labelstr $ 300;
    set formats0 (where=(count(variable,"#")=1 and
index(variable, "VTST#") > 0));
    *get the first indexed # location;
    loc=index(variable,"#");
    substr(variable, loc, 2) = "&i";
    labelstr=catt(labelstr," #&i");
run;
proc print data=iter&suffix.1 &i (obs=5) noobs;
   var variable labelstr;
run;
%END;
%END;
%MEND DO LIST2;
```

# **Practical applications**

## Variable Labels

The iteration techniques discussed above are employed in several different scenarios: data quality checks, creating variable labels, creating format assignment statements, driving range checks, and producing missingness reports. Below follow snippets of code to create a data driven variable label statement. label strings are created by concatenating prefixes and existing variable descriptions using the CATX function, The CATX function is used to add information as a prefix to the label, such as the month of data collection or the tab the variable came from in the data dictionary. The labelstr variable is a sentence that applies a variable label to a variable. When put out to a flat file, it can be included to label variables in a data set.

Assign a filename for the LABEL statement:

```
filename label1 ".\&short. Labels.txt";
```

Create iterations of variables with # signs using macro loops described above:

```
%do_list1(maxiter=3,suffix=id);
%do list2(maxiter=4,suffix=vtst); . . .
```

Add iterated records created by the do loops together:

```
data expand_labels;
    set iterid: itervtst: . . . _ ;
run;
```

Add iterated records to the records that did not require iteration:

```
data labels;
   length variable $ 32;
   set labels0 (where=(index(variable,"#")=0))
        expand_labels (where=(index(variable,"#")=0))
   ;
run;
```

## Output the LABEL statement:

# Include the LABELstatement:

```
filename label1 ".\&short._Labels.txt";
filename retain1 ".\&short. retain.txt";
```

```
run;
data &outfi. (label="Labeled &short");
    retain
        %include retain1;
;
    set &infi.;
    label
        %include label1;
;
run;
```

Figure 2 is a snippet of the text file included to produce variable labels.

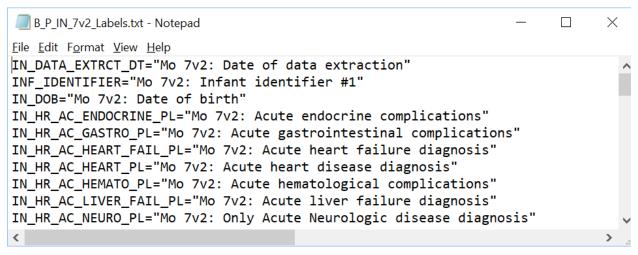
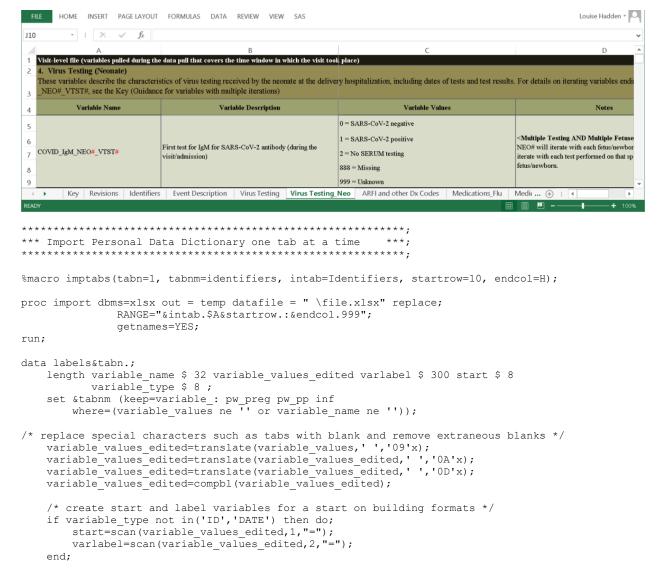


Figure 2. Sample Label Statement Text File

# Value Labels (FORMATS)

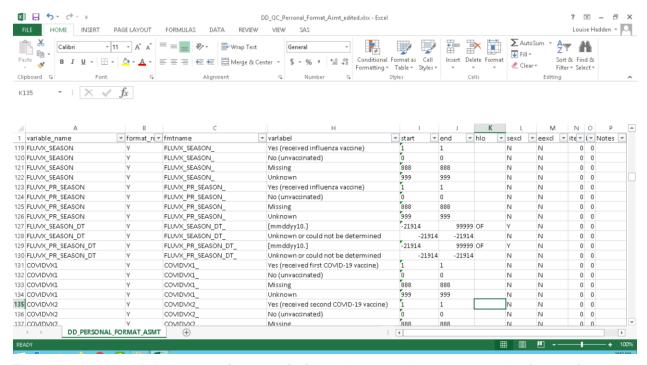
Let us review the data dictionary screenshot again, focusing on the variable values columns and rows. Note that the variable name, variable description and notes fields are single rows, but the value labels are separate rows. Originally, the value labels were on single rows in the Excel data dictionaries but that made parsing the variable values field incredibly difficult so we redesigned the data dictionary to have a separate row for each variable value. When we read in the rows shown, we get five rows of data. Only the first line of data contains variable name, variable description, and notes, while each value label is on a separate row. To produce variable labels, we remove rows in blanks in the variable name field as we saw above, for value labels, we need to fill in those missing cells programmatically.



This is a similar read in as used for variable labels above, but there are key differences. For example, we are keeping records with either the variable name present OR variable value labels present. The reason for this is that some variables do not have formats assigned, or they have a blank line where a range should be specified, so if there is a variable name only, we keep the record. This will be addressed in a manual review step later, and if required, the data dictionary corrected.

In the code snippet below, we fill in the missing rows with the RETAIN statement, and create a format name, among other things. We then export the temporary data set to a spreadsheet for manual checks and adjustments.

```
retain _variable_type;
if not missing(variable_type) then _variable_type=variable_type;
else variable_type=_variable_type;
drop _variable_type;
formatstr=variable_values_edited;
if variable name ne '' then fmtname=cats(variable name,' ');
```



The screenshot above shows some (but not all) of the columns used to create complex formats from a data set. This spreadsheet output is reviewed carefully, any corrections made, and then it is imported into a SAS data set for use in creating formats and assignment statements.

# Building a format library programmatically

We are keeping variable name in our SAS data set derived from the spreadsheet above, so that we can create FORMAT statements as well as a format catalog. For expediency, we name the format name with the variable name with a trailing underscore, stripping the iterator pound signs. To build a library, you need the following three fields:

FMTNAME - format name

LABEL - value label

START - start of a range or value

Additional fields that are used in our processing are:

Variable\_name – used to build format assignment statements

Format\_required – some variables do not require a format

END – end of a range

HLO - specialized formats - high, low, other

SEXCL (exclude the start of the range)

EEXCL (exclude the end of the range)

We have some complex formats for dates, ranges and nested formats which require END, HLO, SEXCL and EEXCL.

#### How do FORMATS work?

The best way to figure out how formats work is to analyze them. The client for this project wanted to assign special date values for missing values. We create a small data set with the special dates and explore to see how this complex format looks in various forms. The same technique can be used to look

at ranges. What we are trying to achieve is an input data set which looks like what SAS expects under all conditions.

```
data temp;
  d1='01jan1900'd; d2='01jan1960'd; d3=today(); d4='01jan1940'd;
proc print data=temp;
run;
proc print data=temp;
format d1 d2 d3 d4 mmddyy10.;
proc format fmtlib;
     value foo '01jan1900'd='Invalid'
               '01jan1940'd='Still in'
               '01jan1960'd='SAS zero'
               other=[mmddyy10.];
run;
proc print data=temp;
format d1 d2 d3 d4 foo.;
proc format cntlout=foo2;
proc print data=foo2;
run;
```

Below we see the number representation of the special dates, followed by their formatted version (SAS data format), followed by our user-defined format.

```
Obs
       d1
              d2
                     d3
 1
       -21914
                   22475
                             -7305
                    d2
                                43
       01/01/1900
                 01/01/1960
                             07/14/2021
                                            01/01/1940
Obs
                  d2
                             d3
                                         d4
 1
      Invalid
                SAS zero 07/14/2021
                                       Still in
```

Below follows the result of PROC FMTLIB, showing how SAS represents the special date format in printed form. Note the other – this is a nested format indicating that any "other" dates should appear in MMDDYY10. format. You can use any SAS-supplied or user-created format as long as the program has access to where the format is stored.

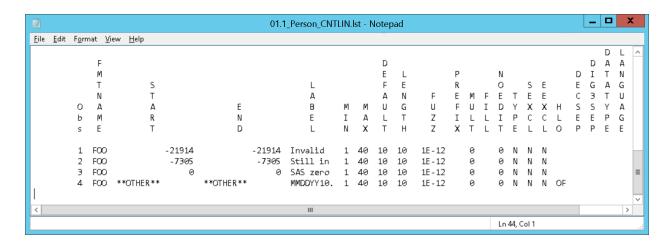
PROC FORMAT CNTLOUT produces a SAS data set from a format catalog file, which produces yet another vision of the same format. It is this version that we need to reproduce in order to use metadata to create format catalogs.

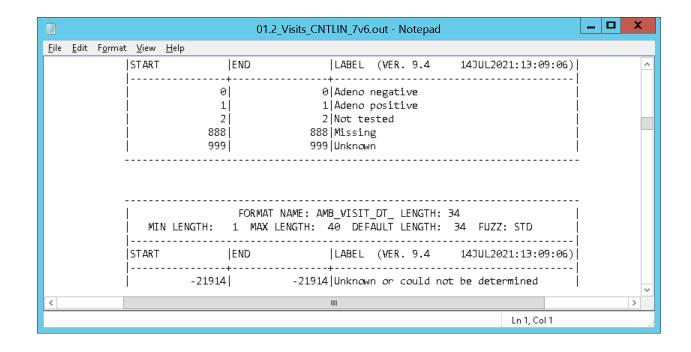
# Using PROC FORMAT CNTLIN

Here we create the input to PROC FORMAT CNTLIN from our data dictionary import file, making adjustments to conform to SAS' requirements for CNTLIN data sets.

```
01.1_Person_CNTLIN.sas - Notepad
File Edit Format View Help
*** Create CNTLIN file
data personformats &procmo (keep=variable name fmtname start end
    label type hlo sexcl eexcl iterated);
    length fmtname $32 type $1 start $14 label $300;
    set person_cntlin (where=(format_req ne 'N') rename=(varlabel=label));
    if variable_type in('CHAR','ID') then type='c';
    else type='n';
    iterated=(index(variable name, '#')>0);
label fmtname = 'Name of Format'
      start = 'Start of Range for Format'
      end = 'End of Range for Format'
      sexcl = 'Starting value excluded from Range'
      eexcl = 'Ending value excluded from Range'
          label = 'Label for Format'
                                                                                              ≣
          type = 'Type of Format'
          hlo = 'High-Low-Other flag'
          iterated = 'Iterated variable';
run;
                                                                                            >
                                                                         Ln 151, Col 27
```

proc format library=library.personformats cntlin=personformats\_&procmo fmtlib; run;

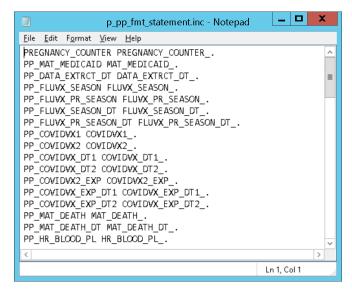




# Creating a format assignment statement programmatically

```
data fmtstm;
  length fmtstm $ 80 fmtdot $ 33 variable_name $ 32;
  file ".\&outfi._fmt_statement.inc" lrecl=80;
  set temp (where=(indata ne 0));
  fmtdot=cats(fmtname,'.');
  fmtstm=catx(' ',variable_name,fmtdot);
  put fmtstm;
run:
```

We use the same data set used to create the format catalogs / SAS data sets containing formats to generate format assignment statements. Note the use of SAS functions to add the dot following the format name, as this does not exist in the SAS data set. This format statement can be included in programs to analyze and characterize the data files. Below follows a screenshot of the format assignment file.



#### **SCENARIO 2**

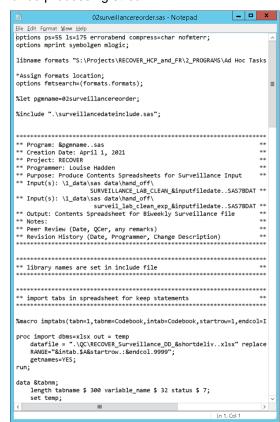
Our second project is also a large scale, multi-site CDC study involving surveillance, COVID and flu testing, and other health metrics including lab data handled separately. Data dictionaries are used extensively, but unlike scenario 1, we do not start with set data dictionaries. The REDCap system is highly flexible and allows for the addition of new forms over time, which leads to different data dictionaries, variable lists, etc. Further, different sites may deploy REDCap forms at a slightly different pace, resulting in input streams with different variables, variable types, etc. REDCaps for each site generate data dictionaries, format programs, and csv files for each collection of forms each week, and these are reviewed for the following weeks' processing. Keeping up with the changes is a full time job for several people managing different processing streams. It is not in the scope of this particular paper to drill down on the exact processes used to manage the task of weekly file generation for the CDC, but we will discuss how we use data dictionaries to design a system to document files sent to the CDC, and compare data and documentation from REDCap (and more) with the final data and documentation sent to CDC. We will focus on a single data stream, surveillance data.

Solutions include using data management tools such as control files, and the use of SAS procedures such as PROC CONTENTS, ODS OUTPUT, and PROC COMPARE. We produce preliminary metadata on incoming weekly surveillance data, which has been pre-processed in a separate set of steps. A control data dictionary is produced from the prior week's delivery, and includes any changes or additions made during the prior week based on review of added forms, etc. The "curated" metadata is used to create a draft deliverable file (corrected position, etc.), and new metadata is produced. Metadata from the incoming file and the curated metadata is compared, and additionally, the format library created by the incoming REDCap data stream is compared to the curated format library. This allows the detection of anomalous data and inconsistent formatting.

# Using a control file

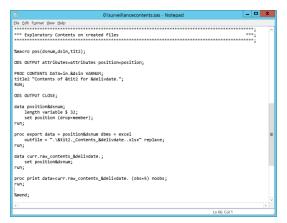
A helpful solution for repetitive processes with multiple programs or scripts is the use of a control file. In this case, a week's processing involves hundreds of processes and all the processes use "dated" data. We use a control file for all surveillance processing to set the dates for incoming files and deliveries, and to set up libraries and filenames for processing. This prevents accidental typos or the use of an incorrect file. The date include file is %included in each program in the surveillance processing stream.

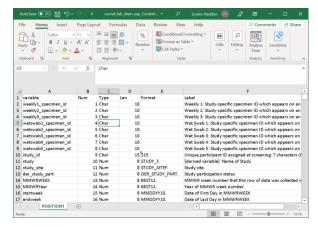




# Create metadata from the incoming surveillance file

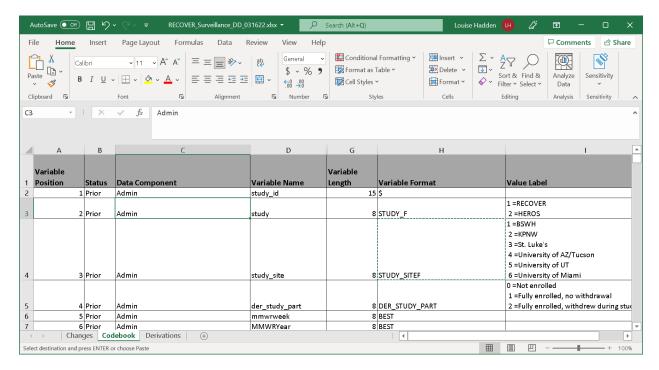
This is a straightforward process. There are myriad ways to report on SAS metadata – Here is one way to produce the elements desired for our process. Note that the results are saved as a permanent SAS dataset.





# Report on metadata from the prior week's surveillance file

The results from the prior week's processing are output in the form of a "codebook" which includes both contents information from the prior week's file and additional items, which include Status, Data Component, and Value Label. Status and Data Component are collected from the prior week's DD, and Value Label is generated programmatically from the format sas data set for the week's processing.



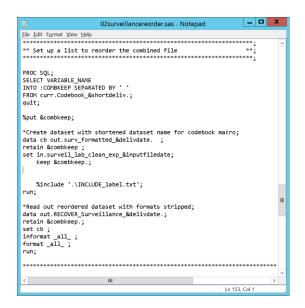
# Importing the prior data dictionary (DD)

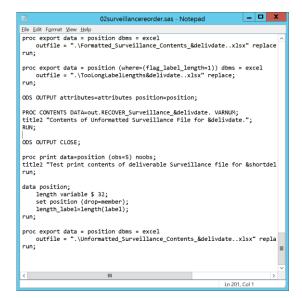
The spreadsheet Codebook tab is read into a SAS data set and saved as a permanent file, via a process which takes metadata from the current data, the prior "curated" DD, and a module which takes the result of looking at the format catalog and linking value labels to variable names. The first code snippet below

imports the codebook, and the second generates include files (text) for use in creating the deliverable sas data set, including the status, data component, and label fields.

# Reordering the incoming data set

You may have noticed that the contents of the incoming file had a different variable order than the DD from the prior delivery. This piece of the program reorders the variables to match what is desired, and exports the contents of the deliverable file.

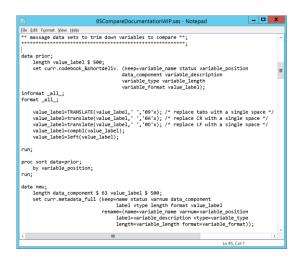


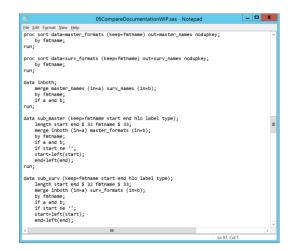


# Preparing the metadata files for comparison by cleaning and standardizing

We are skipping two programs – one is to create a weekly format library and another is to create a new DD in the identical format as the prior DD. As you have seen, we have created permanent metadata files of steps along the way. First up, we try to give the metadata a fair chance by making it as equivalent as

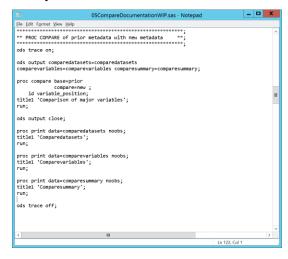
possible. Anything someone has edited in Excel has a good chance of having garbage in it like tabs and line feeds so we strip them.

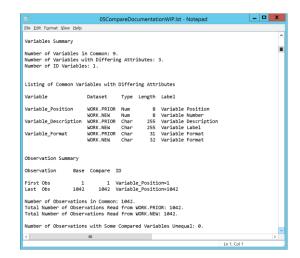




We move on to a comparison of the two metadata files, and then the two format catalogs in the form of SAS data sets.

# **Comparison of Metadata files**





# Comparison of format catalogs

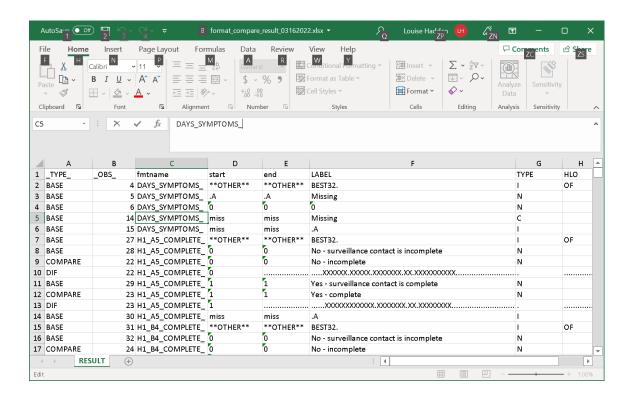
Below follows a format comparison – as you can see, the master format catalog, which is updated weekly but retains error-laden value labels, has coded missing values, changes in value labels, etc. For those of you who are not familiar with PROC COMPARE output, any mismatches in presence (in this case value labels that are only in one format catalog or the other) are output with an indicator of whether they are in the BASE or COMPARE data set. Any mismatches in data values have three lines – the BASE record, the COMPARE record, and a DIFF record. The DIFF record has a "mask" which indicates where differences are found in the variable value. In our case, we see that some value labels have changed over time, and we need to address to provide the client with consistent data.

PROC COMPARE requires the files to be sorted when using the ID statement. We want to match on the format name and start value. Rather than rely on printed output, we use the OUT= option of PROC COMPARE to output the comparison and then export to a spreadsheet.

# Code snippet of PROC COMPARE code

```
X
                      05CompareDocumentation.sas - Notepad
File Edit Format View Help
proc sort data=sub_master;
    by fmtname start;
run;
proc sort data=sub_surv;
    by fmtname start;
run;
proc compare base=sub_master
             compare=sub_surv out=result outnoequal outbase outcomp outdif
     noprint;
    id fmtname start;
run;
proc print data=result noobs;
title1 'Comparison of master and surveillance formats';
run;
<
                        Ш
                                                                              >
                                                               Ln 1, Col 1
```

# Spreadsheet of format comparison output



# **CONCLUSION**

It is clear that using metadata to drive processing and quality assurance can be incredibly helpful and comprises a valuable addition to one's SAS toolbox, along with PROC FORMAT and PROC COMPARE. I hope you will have some fun iterations with functions with your metadata as well.

# **ACKNOWLEDGEMENTS**

This type of complex programming and processing is a team sport. I could not have created and implemented these techniques on my own. A very heartfelt thank you to all my Abt colleagues working on COVID projects.

# **CONTACT INFORMATION**

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