



# EFFICIENT TLF REVIEW WITHIN STATISTICAL PROGRAMMING USING R AND SAS

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1

TRADITIONAL QC PROCESS

5

EXPLORING R PACKAGES

2

CHALLENGES & GAPS

6

R PROGRAMMING SCENARIO

3

RTF REVIEW & SELECTIVE QC

7

R SHINY

4

SAS PROGRAMMING USE CASE

8

SUMMARY

# VALIDATION OF TLF RTF OUTPUT

## QC PROCESS

- Traditionally, all outputs undergo independent QC through double programming.
- Involves double programming of the final SAS dataset.
- Manual review of RTF outputs for any Title, footnotes, page break, and formatting issues.
- There are studies where full independent qc is not required but a targeted QC approach is followed by performing a quality check by reviewing the outputs directly.

## GAPS

- While effective for data verification, PROC COMPARE fails to catch formatting issues like misaligned text, incorrect font styles, or missing footnotes.
- These issues in RTF outputs would not be detected by PROC COMPARE, leading to quality concerns from reviewers.
- Constrained by resource availability.
- Manual review of over 300 RTF outputs is time-consuming and error-prone.

## EFFICIENT TFL REVIEW ENHANCING QUALITY

- This presentation proposes an efficient solution using R and SAS to streamline the review process, ensuring accuracy and consistency across Tables, Listings, and Figures (TLF) package
- The programmatic workflow includes: RTF File Processing:-
- Extracting and analyzing text from RTF files
- Generating reports to highlight inconsistencies
- By optimizing these tasks within Statistical Programming Team, we significantly reduce review time, enhance accuracy, and ensure the TLF package is robust and ready to share with Sponsors and cross-functional Team

# CHALLENGES

## TIME-CONSUMING PROCESS ERROR PRONE VERIFICATION

- Manually reviewing hundreds of RTF outputs requires significant time and resources, creating bottlenecks in tight submission timelines
- Reviewers must check population counts, row consistency, decimal precision, alignment, spelling, titles, footnotes, and page breaks across numerous files

## SELECTIVE QC

- Population counts – Big N
- Primary and Secondary endpoints counts
- Adverse Events – Overall AE, Serious AE, Related AE
- Death
- Decimal precision and alignment
- Page Breaks
- Title and Footnotes
- Spelling Check
- Formatting

## EFFICIENT TFL REVIEW ENHANCING QUALITY

- The entire review process need to be streamlined using R and SAS
- Both SAS and R have capabilities to read RTF files
- Parsing and processing these datasets creates a robust programmatic review tool
- This tool can be used as a post-PROC COMPARE stage to perform a sanity check
- Results from the programmatic review would be output to an Excel file or a Smartsheet dashboard



## CLINICAL REPORTING R PACKAGES

- R packages that are predominantly used in Clinical Reporting as cataloged on pharmaverse
- This list from pharmaverse.org encapsulates all the open source R packages (and code bases from other open source languages) that fall under the **scope of usage for end-to-end clinical reporting in pharma**.
- Pharmaverse is now a PHUSE Working Group



Population Count (Big N)

Primary & Secondary endpoints

Adverse Events

Serious Adverse Event

Related Adverse Event

Decimal Precision & alignment

Titles and Footnotes

Formatting, Spell Checks

Demographic and Baseline Characteristics

Unresolved macro variables

# R PACKAGES FOR GENERATION OF TFL

## Tables



**rtables**   CRAN 0.6.11 Stars 231 Contributors 28  
A framework for declaring complex multi-level tabulations and then applying them to data



**chevron**   CRAN 0.2.10 Stars 12 Contributors 28  
Holds TLG template standards to create standard outputs for clinical trials reporting with limited parameterisation.




**pharmaRTF**   CRAN 0.1.4 Stars 33 Contributors 3  
Enhanced RTF wrapper written in R for use with existing R tables packages such as huxtable or GT



**Tplyr**   CRAN 1.2.1 Stars 95 Contributors 7  
To simplify the data manipulation necessary to create clinical reports





**gtsummary**   CRAN 2.1.0 Stars 1086 Contributors 40  
Creates tables from either an Analysis Results Dataset (ARD) or a data frame with an ARD by-product



**cards**   CRAN 0.5.0 Stars 39 Contributors 14  
Construct CDISC Analysis Results Dataset (ARD) objects



**cardx**   CRAN 0.2.3 Stars 19 Contributors 16  
Extra Analysis Results Data (ARD) summary objects supplementary to {cards}



**tfrmt**   CRAN 0.1.3 Stars 72 Contributors 10  
A language for defining display-related metadata to automate the transformation from an Analysis Results Dataset (ARD) to a table





**tfrmtbuilder**   CRAN 0.1.0 Stars 5 Contributors 2  
A Shiny app interface for the {tfrmt} package



**tidytlg**   CRAN 0.1.5 Stars 33 Contributors 7  
Generate table, listings, and graphs (TLG) using the Tidyverse

## Listings



**rlistings**   CRAN 0.2.10 Stars 27 Contributors 19  
A framework for creating data listings



## Plots



**ggsurvfit**   CRAN 1.1.0 Stars 76 Contributors 8  
Eases the creation of time-to-event (aka survival) summary figures



While ggplot2 is a lower level, non-pharma specific plotting package. It is universally accepted as the package for graphics, so included here and as a non-pharma package.





**ggplot2**    
An implementation of the Grammar of Graphics in R, and the most popular plotting package for static plots in R.

## Interactive



**tidyCDISC**   CRAN 0.2.1 Stars 106 Contributors 8  
A shiny app to easily create custom tables and figures from ADaM-ish data sets



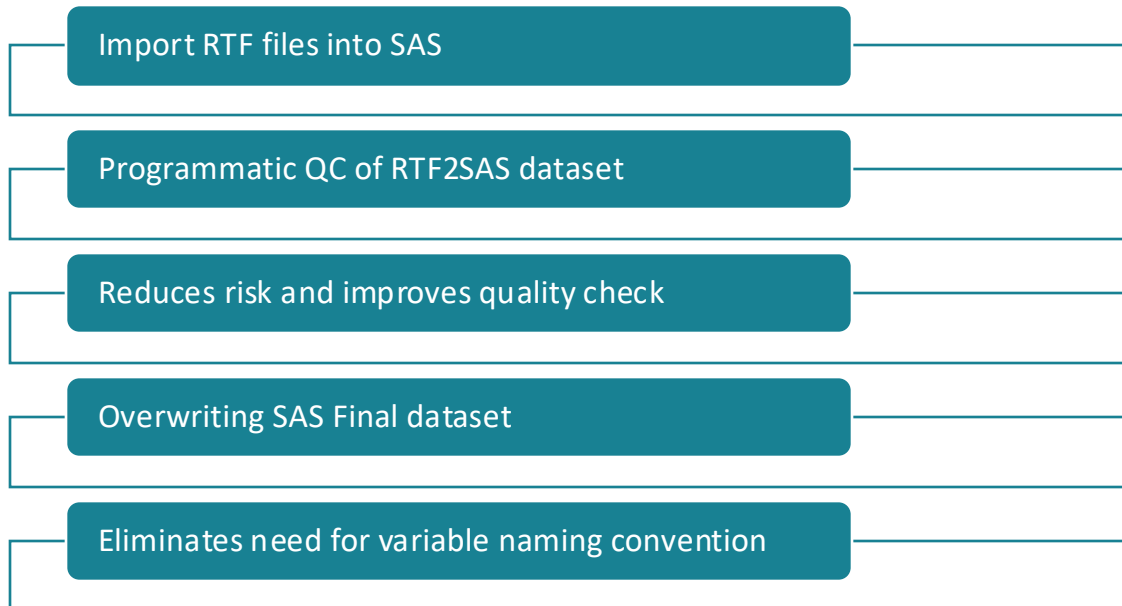
**rhino**   CRAN 1.10.1 Stars 303 Contributors 19  
Supports creating and extending enterprise Shiny applications using best practice

## PHARMAVERSE

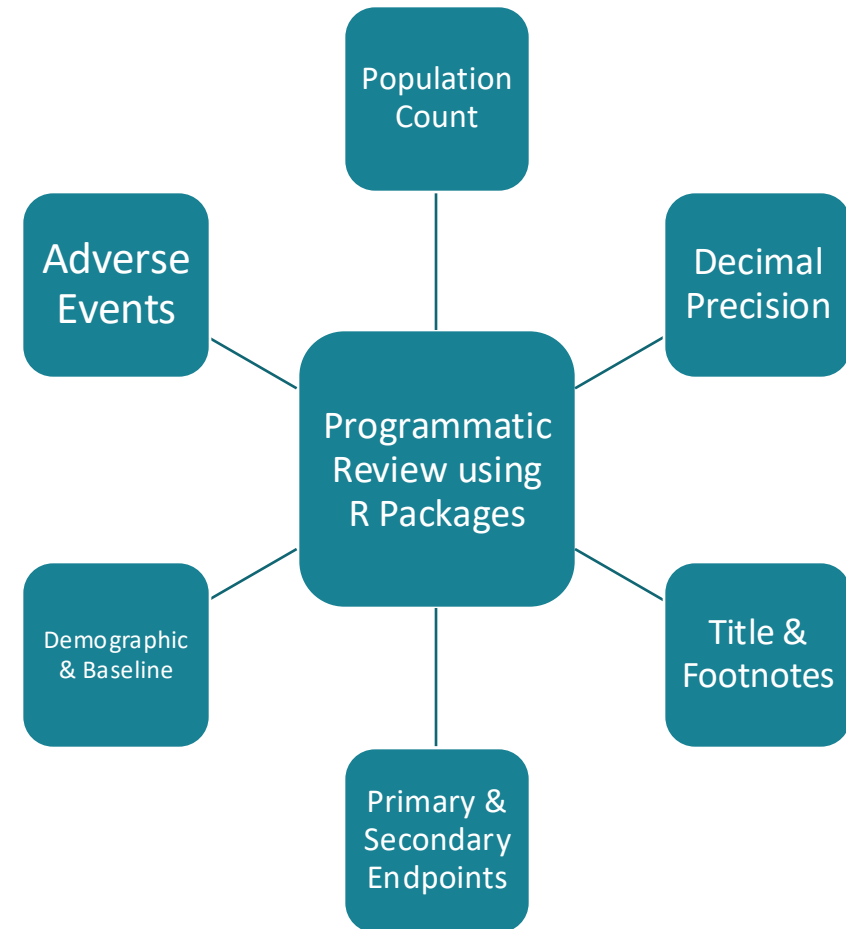
→ R packages commonly used for **generation** of Tables, Listings, and Figures, as outlined on Pharmaverse

# PROPOSED PROGRAMMATIC QC PROCESS

## Step 1



## Step 2



# STEP 1

## READ RTF FILE INTO SAS DATASET

```
/*read in the RTF dataset*/
data readrtf;
infile "&rtf./&outname..rtf" missover length = 1 end = lastobs lrecl = 2000;
input string $varying2000. 1;
rownum = _n_;
run;

/*set read in RTF data and assign values to data*/
data rdl;
set readrtf;
by rownum ;
retain trow rrow page 0;
if index(string, '\headery') then page=page+1; *assign number of page base
if index(string, '\trowd') then trow=trow+1; *assign trow based upon trc
if index(string, '\row') then rrow=rrow+1; *assign rrow based upon act

run;
```

- Extract data from RTF file
- Parse data
- Programmatic Check for Counts for Key demographic, Primary and Secondary endpoints

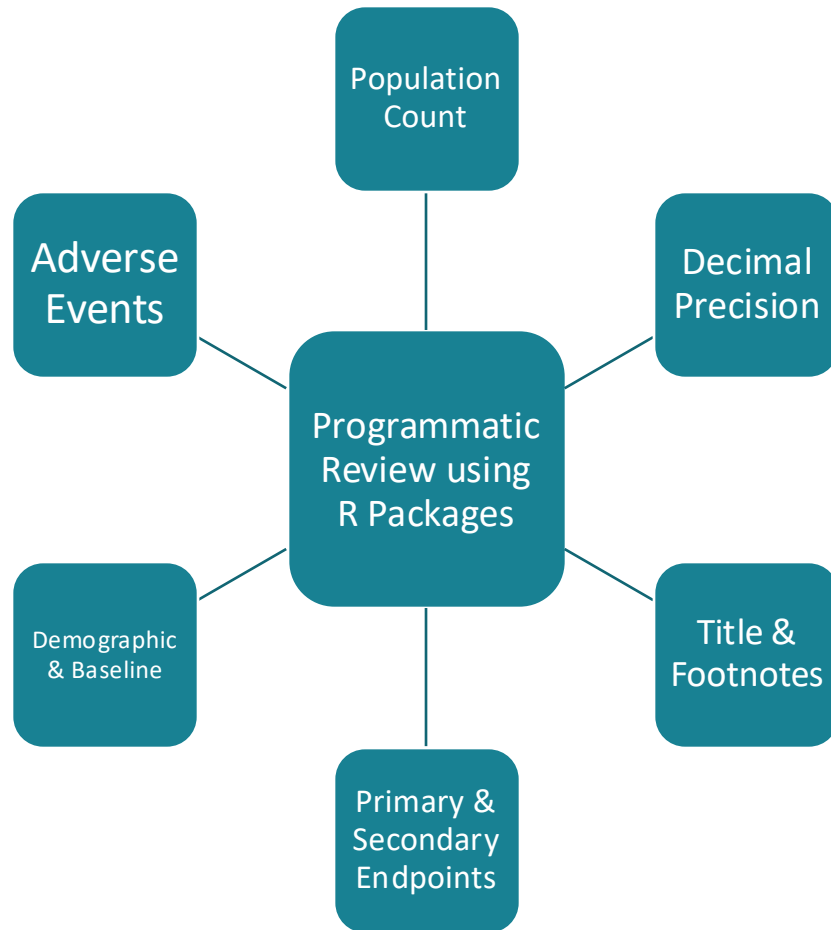
## SAS OUTPUT DATASET

	A1	A2	A3	A4	A5	OBS	#
10	Sexn(%)	n	100	91	191	10	
11		Male	62(62.0)	52(57.1)	114(59.7)	11	
12		Female	38(38.0)	39(42.9)	77(40.3)	12	
13	Femalechildbearingpotentialn(%)	n	38	39	77	13	
14		No	37(97.4)	39(100.0)	76(98.7)	14	
15		Yes	1(2.6)	0	1(1.3)	15	
16	Racen(%)	n	100	91	191	16	
17		BlackorAfricanAmerican	1(1.0)	1(1.1)	2(1.0)	17	
18		White	99(99.0)	89(97.8)	188(98.4)	18	
19		Other	0	1(1.1)	1(0.5)	19	
20	Ethnicityn(%)	n	100	91	191	20	
21		Hispanic/Latino	1(1.0)	1(1.1)	2(1.0)	21	
22		NotHispanic/Latino	93(93.0)	89(97.8)	182(95.3)	22	
23		Notreported	6(6.0)	1(1.1)	7(3.7)	23	



# R PACKAGES USED FOR SELECTIVE QC

## STEP 2



Shiny	• Build interactive web applications with R
Striprtf	• Extract Text from RTF File
Stringr	• Consistent wrapper for common string operations
dplyr	• Tool for working with data frame like objects
writexl	• Export data frames to excel 'xlsx' format
gtsummary	• Creates table summarizing datasets, regression models
Hunspell	• Spell checker

# REVIEW OF RTF OUTPUTS

Paper ET02  
EMERGING TRENDS  
PHUSE US Connect 2025

Table 14.PHUSE  
Summary of Participant Global Impression of ~~Change~~ by Visit - Double-Blind Phase  
Intent-to-Treat Population

1-	Treatment (N=71)	Placebo (N=101)
Visit		
Visit 2		
Very Much Better	6 (10%)	2 (4%)
Moderately Better	13 (22%)	3 (6%)
A Little Better	14 (24%)	4 (8%)
No <del>Change</del>	10 (17%)	28 (55%)

Table 14.1.2.1.1  
Demographics and Other Baseline Characteristics - Double-Blind Phase  
Intent-to-Treat Population

	Total (N=142)
Sex, n (%)	
Male	50 (35%)
Female	92 (65%)
Race, n (%)	
White	118 (83%)
Black or African American	9 (6%)
Asian	
Baseline	
n	56
Mean (SD)	86.3 (16.51)
Median	85.0
Min, Max	57, 132
Screening	
n	56
Mean (SD)	87.1 (15.85)
Median	85.5
Min, Max	55, 133

Database Extraction: 17-Mar-2025  
CTCAE = Common Terminology Criteria for Adverse Events  
Note: Relative Day is the day relative to the first dose of study medication, Day 1.  
Note: \* indicates the assessment was completed out of window.  
[1] Flag is in relation to the normal range (ex. High, Low, Normal).  
Source: P:\l\_lab\_phuse.sas, Date/time of run: 17MAR2025:12:02

Page 10 of 2554

## RTF REVIEW

Population Count (Big N)

Primary & Secondary endpoints

Adverse Events

Serious Adverse Event

Related Adverse Event

Decimal Precision & alignment

Titles and Footnotes

Formatting, Spell Checks

Demographic and Baseline Characteristics

Unresolved macro variables

## READ & EXTRACT USING R PACKAGES

# STRIPRTF - EXTRACT TEXT FROM RTF FILE

```
install.packages("stripdf")
library(stripdf)

rtf_text1 <- read_rtf ("C:/Users/kumars/Documents/TLF/t_pgic_
rtf_text1

> rtf_text1 <- read_rtf ("C:/Users/kumars/Documents/TLF/t_pgic_phuse.rtf")
> rtf_text1
[1] "*" | "
[2] "*" | Visit | | Treatment\n(N=71) | Placebo\n(N=101) | "
[3] "*" | "
[4] "*" | Visit 2 | | | "
[5] "*" | Very Much Better | | 6 ( 10%) | 2 ( 4%) | "
[6] "*" | Moderately Better | | 13 ( 22%) | 3 ( 6%) | "
[7] "*" | A Little Better | | 14 ( 24%) | 4 ( 8%) | "
[8] "*" | No Chagne | | 10 ( 17%) | 28 ( 55%) | "
[9] "*" | A Little worse | | 10 ( 17%) | 12 ( 24%) | "
[10] "*" | Moderately worse | | 3 ( 5%) | 1 ( 2%) | "
[11] "*" | Very Much worse | | 2 ( 3%) | 1 ( 2%) | "
[12] "*" | "
[13] "*" | Visit 3 | | | "
[14] "*" | Very Much Better | | 10 ( 19%) | 1 ( 2%) | "
[15] "*" | Moderately Better | | 13 ( 25%) | 5 ( 8%) | "
[16] "*" | A Little Better | | 8 ( 15%) | 12 ( 20%) | "
[17] "*" | No Chagne | | 10 ( 19%) | 28 ( 47%) | "
[18] "*" | A Little worse | | 9 ( 17%) | 9 ( 15%) | "
```

# UNRTF - CONVERTS AN RTF DOCUMENT TO HTML, TEXT

[illegible]

# SCENARIO 1 – POPULATION COUNTS

## BIG N

- Population counts (Big N) are assigned during the PROC Report stage in SAS
- Potential Issue : Incorrect calculation of values
- Incorrect assignment of values, such as placebo counts appearing in the treatment group and vice versa.
- Automation Using R: R packages can extract Big N values from the RTF outputs to ensure accurate population counts.

```
library(stringr)
library(striprtf)
library(dplyr)

rtf_text <- read_rtf ("C:/Users/kumars/Documents/TLF/AP/t_14_1_1_1_disp.rtf")
rtf_text ## Character Vector

extracted_T1 <- str_extract(rtf_text, "Treatment[^)]*\\)")
extracted_T1

extracted_T2 <- str_extract(rtf_text, "Placebo[^)]*\\)")
extracted_T2

> rtf_text <- read_rtf ("C:/Users/kumars/Documents/TLF/t_pgic_phuse.rtf")
> rtf_text
[1] "*" | "
[2] "*" | Visit | | Treatment\\n(N=71) | Placebo\\n(N=101) | "
[3] "*" | "
[4] "*" | Visit 2 | | | "
[5] "*" | Very Much Better | | 6 ( 10%) | 2 ( 4%) | "
[6] "*" | Moderately Better | | 13 ( 22%) | 3 ( 6%) | "
[7] "*" | A Little Better | | 14 ( 24%) | 4 ( 8%) | "
```

	pages	extracted_T1	extracted_T2
1	13	Treatment (N=71)	Placebo (N=101)



...	...6	...11	...12	Cc	rtfnan	nam	Treatment	Table	Placebo
Table	t_14_1_2_1	Table 14.1.2.	Demographics and Other Baseline Chara	C2	F:/Biome	c2	Treatment (N= 71)	c2	Placebo (N=100)
Table	t_14_2_1_5	Table 14.2.1.	Progression Free Survival by Subgroups -	C20	F:/Biome	c20	Treatment (N= 71)	c20	Placebo (N=100)
Table	t_14_2_2_2	Table 14.2.2.	Objective Response Rate by Subgroups -	C22	F:/Biome	c22	Treatment (N= 71)	c22	Placebo (N=100)
Table	t_14_2_13	Table 14.2.13	Summary of Participant Global Impressio	C57	F:/Biome	c57	Treatment (N= 71)	c57	Placebo (N=100)
Table	t_14_2_14	Table 14.2.14	Summary of Participant Global Impressio	C58	F:/Biome	c58	Treatment (N= 71)	c58	Placebo (N=100)
Table	t_14_3_1_1	Table 14.3.1.	Overall Summary of Treatment-Emerger	C59	F:/Biome	c59	Treatment (N= 71)	c59	Placebo (N=100)
Table	t_14_3_1_2	Table 14.3.1.	Treatment-Emergent Adverse Events by	C60	F:/Biome	c60	Treatment (N= 71)	c60	Placebo (N=100)
Table	t_14_3_1_3	Table 14.3.1.	Treatment-Emergent Adverse Events Re	C61	F:/Biome	c61	Treatment (N= 71)	c61	Placebo (N=100)
Table	t_14_3_1_4	Table 14.3.1.	Serious Treatment-Emergent Adverse Ev	C62	F:/Biome	c62	Treatment (N= 71)	c62	Placebo (N=100)
Table	t_14_3_1_5	Table 14.3.1.	Serious Treatment-Emergent Adverse Ev	C63	F:/Biome	c63	Treatment (N= 71)	c63	Placebo (N=100)
Table	t_14_3_1_6	Table 14.3.1.	Treatment-Emergent Adverse Events wi	C64	F:/Biome	c64	Treatment (N= 71)	c64	Placebo (N=100)
Table	t_14_3_1_2	Table 14.3.1.	Treatment-Emergent Adverse Events by	C79	F:/Biome	c79	Treatment (N= 71)	c79	Placebo (N=100)
Table	t_14_3_1_2	Table 14.3.1.	Serious Treatment-Emergent Adverse Ev	C80	F:/Biome	c80	Treatment (N= 71)	c80	Placebo (N=100)

# R SHINY

```
library(shiny)
library(dplyr)
library(readxl)
library(writexl)
library(striprtf)
library(stringr)

ui <- fluidPage(
  titlePanel("RTF Population (Big N) Extractor"),
  sidebarLayout(
    sidebarPanel(
      fileInput("folder", "Select Folder Containing RTF Files", multiple = TRUE, accept = ".rtf"),
      textInput("treatment1", "Enter Treatment 1 Name", value = "Treatment"),
      textInput("treatment2", "Enter Treatment 2 Name", value = "Placebo"),
      downloadButton("downloadData", "Download Excel")
    ),
    mainPanel(
      tableOutput("extractedData")
    )
  )
)

server <- function(input, output) {
  extract_big_n <- function(file_path, treatment1, treatment2) {
    rtf_text <- read_rtf(file_path)
    extracted_T1 <- na.omit(str_extract(rtf_text, paste0(treatment1, "[^]*\\s")))
    extracted_T2 <- na.omit(str_extract(rtf_text, paste0(treatment2, "[^]*\\s")))
    data.frame(File = basename(file_path), T1 = ifelse(length(extracted_T1) > 0, extracted_T1[1], NA), T2 = ifelse(length(extracted_T2) > 0, extracted_T2[1], NA))
  }
}
```

## RTF Population (Big N) Extractor

Enter Folder Path Containing RTF Files

Enter Treatment 1 Name

Enter Treatment 2 Name

 Download Excel

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# SCENARIO 2 – DECIMAL PRECISION

## COUNTS

- Ensuring decimal consistency across all outputs, as defined in the SAP, is essential.
- The lead programmer is responsible for verifying that all outputs adhere to the specified decimal precision.
- The program captures actual decimal precision in outputs, compares it against SAP-defined precision, and flags discrepancies.
- The R package Striprtf is used to read RTF files, with further processing using stringr, dplyr, dt, and Shiny.
- The Shiny interface allows users to input file locations and review flagged inconsistencies efficiently.



## RTF Decimal Precision Checker

Enter Folder Path:

Check Files
Download Results

Show 10 entries
Search:

	FileName	Content	ConsistentDecimalPlaces
318	t_14_3_2_4_1_eg_visit.rtf	*  Mean (SD)    69.0   69.0	false
324	t_14_3_2_4_1_eg_visit.rtf	*  Mean (SD)    -5.0   -5.0	false
696	t_14_3_2_4_1_eg_visit.rtf	*  Mean (SD)    135.0   135.0	false
702	t_14_3_2_4_1_eg_visit.rtf	*  Mean (SD)    6.0   6.0	false
1074	t_14_3_2_4_1_eg_visit.rtf	*  Mean (SD)    93.0   93.0	false
1080	t_14_3_2_4_1_eg_visit.rtf	*  Mean (SD)    6.0   6.0	false
1452	t_14_3_2_4_1_eg_visit.rtf	*  Mean (SD)    390.0   390.0	false
1458	t_14_3_2_4_1_eg_visit.rtf	*  Mean (SD)    2.0   2.0	false
1830	t_14_3_2_4_1_eg_visit.rtf	*  Mean (SD)    409.0   409.0	false
1836	t_14_3_2_4_1_eg_visit.rtf	*  Mean (SD)    -7.0   -7.0	false

Showing 1 to 10 of 28 entries

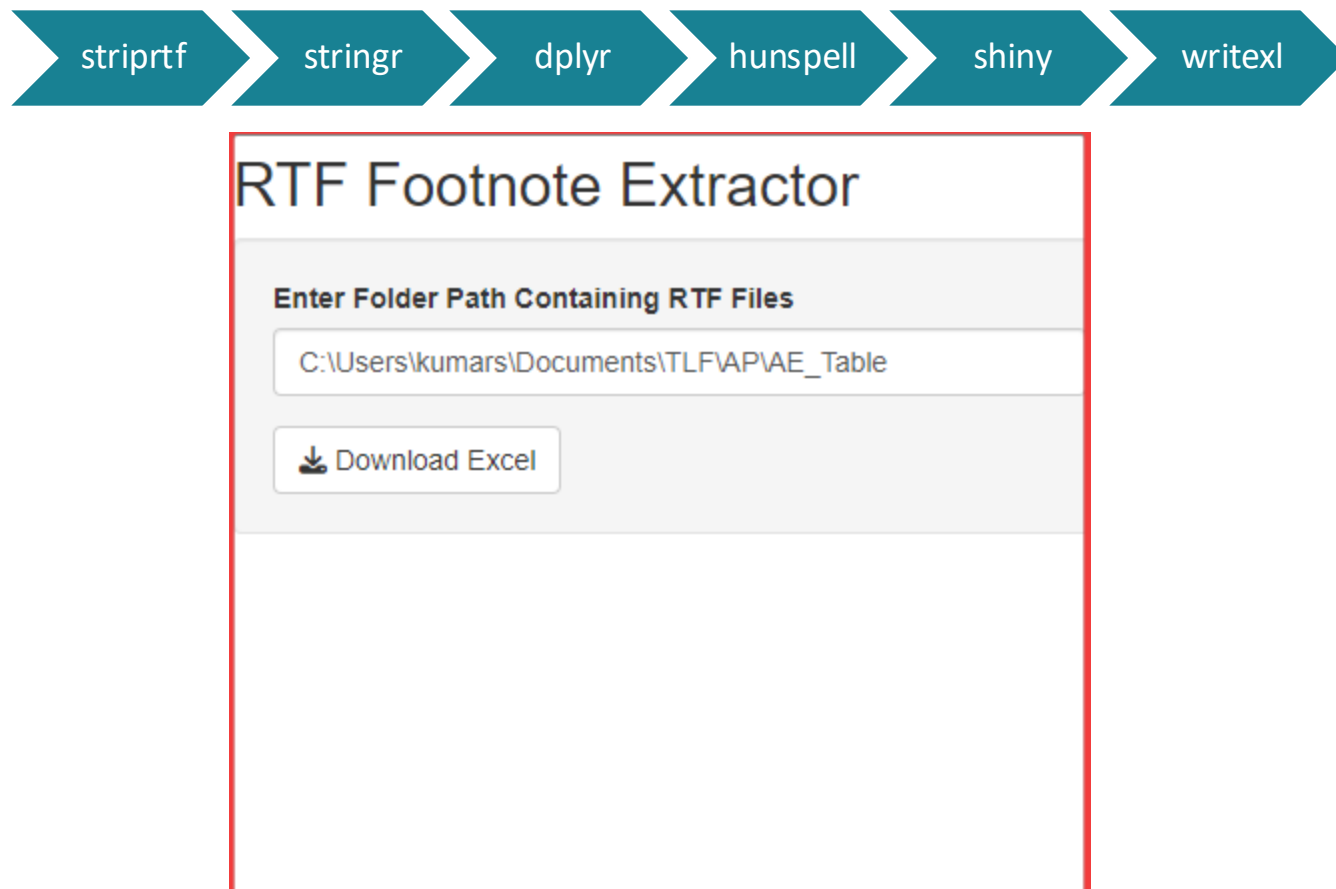
Previous
1
2
3
Next



## SCENARIO 3 – TITLE & FOOTNOTES

### TLF METADATA

- Ensuring that titles and footnotes in RTF outputs are consistent, accurate and error-free
- Identifying discrepancies such as unresolved macros, spell checks, formatting inconsistencies are key during manual review.
- additional R package hunspell used
- Parsing Data – Extract and parse data from the RTF files for further analysis.



# SCENARIO 3 – TITLE & FOOTNOTES



File	Content	SpellCheck
t_14_3_1_1_ae_overall.rtf	AE: Adverse Event; AESI: Adverse Event of Special Interest; CTCAE: Common Terminology Criteria for Adverse Events; MedDRA: Medical Dictionary of Regulatory Activities; NCI: National Cancer Institute;	AE, AESI, CTCAE, MedDRA, NCI
t_14_3_1_1_ae_overall.rtf	TEAE: Treatment-Emergent Adverse Event; SAE: Serious Adverse Event.	TEAE, SAE
t_14_3_1_1_ae_overall.rtf	N represents the number of participants in the corresponding cohort.	No errors
t_14_3_1_1_ae_overall.rtf	All AEs were coded using MedDRA Version 24.0.	AEs, MedDRA
t_14_3_1_1_ae_overall.rtf	The severity of All AEs are graded using NCI-CTCAE Version 5.0: Grade 1=Mild; Grade 2=Moderate; Grade 3=Severe; Grade 4=Life-threatening; Grade 5=Death.	AEs, NCI, CTCAE
t_14_3_1_1_ae_overall.rtf	TEAEs are classified as study drug-related if they have an investigator determination of related to study treatment.	TEAEs
t_14_3_1_1_ae_overall.rtf	A TEAE is defined as an AE that starts or worsens on or after the first dose of study treatment is taken, including those occurring or increasing in severity up to 30 days after the last dose of study treatment.	TEAE, AE
t_14_3_1_1_ae_overall.rtf	A TEAE is defined as leading to a dose modification if the action taken with study treatment for a TEAE was having a dose interruption, dose reduction or dose increase.	TEAE, TEAE
t_14_3_1_1_ae_overall.rtf		No errors

# PYTHON PACKAGE STRIPRTF & HUNSPELL

```
from striprtf.striprtf import rtf_to_text
import re

# Path to your RTF file
file_path = "C:/Users/kumars/Documents/TLF/t_pgic_phuse.rtf"

# Function to read RTF file and extract text
def read_rtf(file_path):
    with open(file_path, 'r', encoding='utf-8') as file:
        rtf_content = file.read()
    return rtf_to_text(rtf_content)

# Process the RTF file
rtf_text = read_rtf(file_path)
rtf_text
```

- Striprtf and hunspell can also be used as python packages.
- Python packages provides capabilities to output to SAS datasets
- Python packages reads rtf and pdf files efficiently

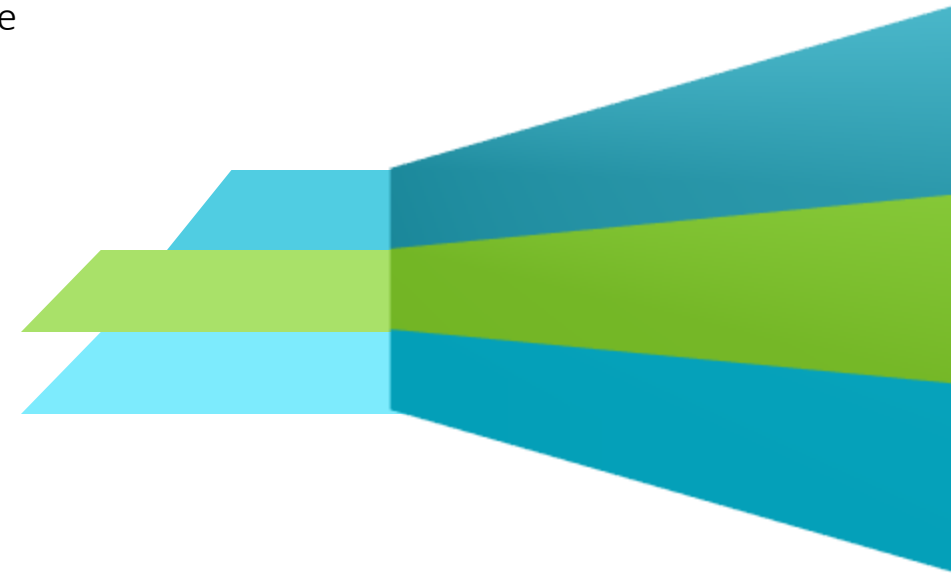
```
|\\nVisit||Treatment\\n(N=71)|Placebo\\n(N=101)|\\n\\nVisit 2||||\\n Very Much Better|| 6 ( 10%)| 2 ( 4%)|\\n Moderately Be
tter|| 13 ( 22%)| 3 ( 6%)|\\n A Little Better|| 14 ( 24%)| 4 ( 8%)|\\n No Chagne|| 10 ( 17%)| 28 ( 55%)|\\n A L
ittle Worse|| 10 ( 17%)| 12 ( 24%)|\\n Moderately Worse|| 3 ( 5%)| 1 ( 2%)|\\n Very Much Worse|| 2 ( 3%)| 1
( 2%)|\\n\\nVisit 3||||\\n Very Much Better|| 10 ( 19%)| 1 ( 2%)|\\n Moderately Better|| 13 ( 25%)| 5 ( 8%)|\\n A L
ittle Better|| 8 ( 15%)| 12 ( 20%)|\\n No Chagne|| 10 ( 19%)| 28 ( 47%)|\\n A Little Worse|| 9 ( 17%)| 9 ( 15%)|
\\n Moderately Worse|| 2 ( 4%)| 2 ( 3%)|\\n Very Much Worse|| 1 ( 2%)| 2 ( 3%)|\\n\\nVisit 4||||\\n Very Much Be
tter|| 14 ( 31%)| 0|\\n Moderately Better|| 6 ( 13%)| 2 ( 6%)|\\n A Little Better|| 7 ( 16%)| 6 ( 17%)|\\n No
Chagne|| 7 ( 16%)| 15 ( 43%)|\\n A Little Worse|| 5 ( 11%)| 6 ( 17%)|\\n Moderately Worse|| 6 ( 13%)| 5 ( 14%)
|\\n Very Much Worse\\t|| 0| 1 ( 3%)|\\n\\n\\n\\n\\nVisit||Treatment\\n(N=71)|Placebo\\n(N=101)|\\n\\nVisit 5||||\\n Very Much Bet
ter|| 11 ( 26%)| 2 ( 5%)|\\n Moderately Better|| 9 ( 21%)| 5 ( 12%)|\\n A Little Better|| 5 ( 12%)| 6 ( 15%)|
\\n No Chagne|| 7 ( 17%)| 17 ( 41%)|\\n A Little Worse|| 7 ( 17%)| 4 ( 10%)|\\n Moderately Worse|| 3 ( 7%)| 5
( 12%)|\\n Very Much Worse|| 0| 2 ( 5%)|\\n\\nVisit 6||||\\n Very Much Better|| 7 ( 18%)| 0|\\n Moderately Better||
7 ( 18%)| 1 ( 3%)|\\n A Little Better|| 10 ( 26%)| 7 ( 20%)|\\n No Chagne|| 6 ( 16%)| 15 ( 43%)|\\n A Little Wors
e|| 7 ( 18%)| 9 ( 26%)|\\n Moderately Worse|| 1 ( 3%)| 2 ( 6%)|\\n Very Much Worse|| 0| 1 ( 3%)|\\n\\nVisit
7||||\\n Very Much Better|| 11 ( 26%)| 0|\\n Moderately Better|| 12 ( 29%)| 4 ( 11%)|\\n A Little Better|| 6 ( 1
4%)| 4 ( 11%)|\\n No Chagne|| 9 ( 21%)| 17 ( 49%)|\\n A Little Worse|| 3 ( 7%)| 5 ( 14%)|\\n Moderately Worse||
1 ( 2%)| 4 ( 11%)|\\n Very Much Worse|| 0| 1 ( 3%)|\\n\\n\\n\\n\\nVisit||Treatment\\n(N=71)|Placebo\\n(N=101)|\\n\\nVisit 8|
||\\n Very Much Better|| 8 ( 20%)| 0|\\n Moderately Better|| 8 ( 20%)| 4 ( 14%)|\\n A Little Better|| 9 ( 22%)|
```

# SUMMARY

**Targeted QC** – Efficiently validating key parameters during review and ensuring the accuracy and reliability of TLF outputs.

**Generative AI-powered app could further enhance efficiency.** This process is highly beneficial for sponsors seeking to perform a final sanity check before review and submission. It also serves as a valuable safeguard for CROs, allowing them to conduct a final verification before delivering outputs to sponsors.

Within a statistical programming team, this represents the last quality control checkpoint before the data is shared for cross-functional review, **reinforcing confidence in the integrity of the results.**





# FINAL QUALITY CHECK: ENHANCING ACCURACY, CONFIDENCE, AND EFFICIENCY IN TLF REVIEW

## ENSURING ACCURACY AND RELIABILITY IN TERTIARY QC

Validating key parameters is critical for maintaining the integrity of TLF outputs



## FINAL VERIFICATION FOR SPONSORS AND CROs

A crucial safeguard before submission, enhancing confidence in deliverables



## OPTIMIZING QUALITY CONTROL WITH REVIEW TOOLS

Strengthening data integrity while improving efficiency and reducing errors



# POWERING POSSIBILITIES IN DRUG DEVELOPMENT

Navitas is a knowledge-driven, tech-enabled CRO offering bespoke solutions across the drug development life cycle. From strategic advisory services that navigate complex regulatory landscapes to clinical development ensuring compliant, data-driven operations, and post-marketing solutions enhancing product value with safety surveillance, life cycle management, and real-world evidence—we integrate deep expertise, cutting-edge technology, and industry insights to drive success.



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