# Data Visualisation of Completion Rate for PRO objectives in **Oncology Clinical Trials Supporting PRO Estimands**

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#### OBJECTIVE

- > Longitudinal analysis of PRO data in oncology clinical trials is widely used to assess the effect of treatment on patient-reported quality of life, functioning and disease symptoms.
- > The introduction of the ICH E9(R1) addendum estimand framework and its application in oncology clinical trials has resulted in development of PRO estimands where discontinuation of treatment, disease progression and death are identified as intercurrent events.
- > To align with planned PRO estimands, which may focus on-treatment period or prior to disease progression, it is important to distinguish at each visit PRO data which are:
- **1.** "Expected" and non-missing
- **2.** "Expected" but missing for other reasons
- **3.** "Not expected" (e.g. after disease progression or death events)

#### FDA and SISAQoL Guidance

- > Both SISAQoL recommendations<sup>1</sup> and recent FDA guidance (November 2023)<sup>2</sup> clarify that available data and completion rate should be reported and reasons for missing data should be summarised.
- > The FDA provided illustrations in table and graphical formats proposing presentations of PRO data disposition and completion rate reflecting categories of Expected and Not Expected PRO data

International standards for the analysis of quality-of-life and 🖒 🔜 🌘 patient-reported outcome endpoints in cancer randomised controlled trials: recommendations of the SISAQOL Consortium

# **MSR 36**



Patient-Centered

**Submitting Patient-Reported Outcome Data in Cancer Clinical Trials Guidance for Industry Technical Specifications Document** 

For questions regarding this technical specifications document, contact CDER at cder-edata@fda.hhs.gov



**Technical Specifications Documen** 

#### 5.3.1 Patient Disposition when Evaluating Clinical Benefit

ble A4. Patient Disposition when Evaluating Clinical Benefit (Denominator = Randomized Population

Analysis Visit	Treatment Arm	Randomized Patients (N)	PRO Expected <sup>2</sup>				PRO Not Expected	
			Patients On Therapy, n (%)	Treatment Discontinuation: Disease Progression, n (%)	Treatment Discontinuation: Adverse Event (AE), n (%)	Treatment Discontinuation: Other Reasons, n (%)	Death, n (%)	Other, <sup>3</sup> n (%)
Baseline	Control	600	600 (100.0%)	0 (0.0%)	0 (0.0%)	0 (0.0%)	0 (0.0%)	0 (0.0%)
	Treatment	602	602 (100.0%)	0 (0.0%)	0 (0.0%)	0 (0.0%)	0 (0.0%)	0 (0.0%)
Cycle 2 Day 1	Control	600	564 (94.0%)	16 (2.7%)	15 (2.5%)	0 (0.0%)	5 (0.8%)	0 (0.0%)
	Treatment	602	572 (95.0%)	10 (1.7%)	13 (2.2%)	0 (0.0%)	7 (1.2%)	0 (0.0%)
Cycle 3 Day 1	Control	600	525 (87.5%)	30 (5.0%)	26 (4.3%)	6 (1.0%)	13 (2.2%)	0 (0.0%)
	Treatment	602	542 (90.0%)	23 (3.8%)	21 (3.5%)	0 (0.0%)	16 (2.7%)	0 (0.0%)



> Our goal was to develop a data visualization solution to facilitate more effective interpretation of PRO completion rates and intercurrent events over time.

K.S	A variable denominator rate should be reported.	The normber of patients of FRO assessment identifies those patients who are suffexpected to provide FRO assessments at that threpoint.
19-20	This rate is defined as the number of patients on	Conversely, patients who are off PRO assessments are defined as patients who are no longer expected to provide PRO assessments from that
	PRO assessment submitting a valid PRO	timepoint onwards.
	assessment at the designated timepoint as a	
	proportion of the number of patients on PRO	It was agreed to standardise that PRO assessments after death are considered off PRO assessment and will no longer be included in the
	assessment at the designated timepoint.	denominator of the completion rates (ie, number of patients on PRO assessment). This implicitly implies that unobserved assessments after
	The term completion rate should be used to	death will not be considered as missing data.
	express the rate with the variable denominator	
	rate.	Whether or not to standardise other reasons, such as off PRO protocol, patient withdrawal, and loss to follow-up in the number of patients on
		PRO assessment, needs further discussion (appendix pp 35–36).
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## **DATA VISUALISATION OF COMPLETION RATE: DIVERGENT BAR CHART**

#### **PRO DATA EXPECTED**

Above the line: green bars show "Expected" PRO data so that it is easy to view the completion rate and the amount of missing data in each treatment arm

#### **PRO DATA NOT EXPECTED**

- Below the line, bars in purple reflects data "not expected"
- The dark purple bar below the line shows the cumulative incidence of death over time – which is key in ability to interpret PRO data summaries



The other purple bars below the line indicate visualise the impact of other intercurrent events and that PRO data is "not expected"



## IMPLEMENTATION

- 1) Create a summary of the counts of the number of patients in each category over time based on data in ADQS dataset (ADSL variables which should be copied into ADQS). In this example, PRO data collected whilst on treatment. Therefore, PRO data is not expected after stopping treatment e.g. due to either disease progression or adverse events
- 2) From ADQS, variable PROEXPFL can be used to identify patients at each AVISIT for PRO Expected or PRO Not Expected
- 3) Not expected: Died: The cumulative frequency of deaths over time (by visit) is required: compare DTHDT with nominal AVISIT date to determine deaths at each visit
- 4) Not expected: For each patient at AVISIT when PRO is "not expected" but patient alive use the reasons for treatment discontinuation (DCTREAS)
- 5) Expected: PRO data present or PRO data is missing
- 6) Use the summary of counts in each category. Also possible to present as % of randomized population in each treatment arm (e.g. if 2:1 randomization ratio)

## Implementation in R

#Modify counts to display below x-axis for Not Expected type comp\_rates <- comp\_rates %>% mutate(COUNT\_MOD = ifelse(

grep1('Not Expected:', comp\_rates\$CATEGORY, fixed = TRUE), -1\*comp\_rates\$COUNT, comp\_rates\$COUNT))

This part creates the negative counts for the divergent bar chart

## Implementation in SAS

#### This part adds a -1 for creation of the divergent bar chart

	<pre>/*NOTE 1: Syntax below requires a dataset (D_FIG), by treatment group, counts across completion types over timepoints*/ /*NOTE 2: Need to create formats for your completion types and treatment groups*/ /*1) Prep the dataset*/</pre>
	EDATA D FIG1;
$\searrow$	SET D_FIG;
	/*N_COUNT is your subject count*/
	IF MISSING(N_COUNT) THEN N_COUNT = 0;
	IF CMR_TYPE > 2 THEN N_COUNT = -1 * N_COUNT; /*Creates the Divergent Bar Chart*/
	N_PERCENT = 100*(N_COUNT/N_POP);

#### After setting up your colours and labels, use PROC SGPANEL to create the bar chart

%LET FONT = COURIER; LET FONT SIZE = 8PT; PROC SGPANEL DATA = D\_FIG DATTRMAP = ATTR\_MAP; FORMAT CMR\_TYPE CMR\_TYPE\_F. ARMCD ARMCD F.;

/\*PANEL/PLOT INFO\*/ PANELBY ARMCD / NOVARNAME HEADERATTRS=(SIZE=&FONT\_SIZE. FAMILY="&FONT"); VBAR AVISIT / RESPONSE = N\_PERCENT GROUP = CMR\_TYPE GROUPDISPLAY = STACK GROUPORDER = DESCENDING MISSING NAME = "VBAR" ATTRID = SID;

/\*LEGEND INFO\*, KEYLEGEND "VBAR" / POSITION=BOTTOM TITLEATTRS=(FAMILY="&FONT" SIZE=&FONT\_SIZE.) VALUEATTRS=(FAMILY="&FONT" SIZE=&FONT\_SIZE.);

/\*AXES INFO\*

RUN:

COLAXIS LABEL="Timepoint" VALUEATTRS=(SIZE=&FONT\_SIZE. FAMILY="&FONT") LABELATTRS=(SIZE=&FONT\_SIZE. FAMILY="&FONT") FITPOLICY=ROTATE; ROWAXIS VALUES = ("-100, -80, -60, -40, -20, 0, 20, 40, 60, 80, 100") VALUESDISPLAY=('100' '80' '60' '40' '20' '0' '20' '40' LABEL="Pecent (%) " VALUEATTRS=(SIZE=&FONT SIZE. FAMILY="&FONT") LABELATTRS=(SIZE=&FONT\_SIZE. FAMILY="&FONT"); RUN; QUIT; TITLE; FOOTNOTE;

#Divergent Stacked Bar plot breaks\_values <- seq(-200, 200, by=50)

ggplot (data = comp\_rates, aes(x = AVISIT, y = COUNT\_MOD, fill = CATEGORY)) + geom\_bar(stat = "identity") + fillScale + scale\_x\_discrete(name = ("Visit"), guide = guide\_axis(angle = 45)) + scale\_y\_continuous(name = ("Number of Patients"), breaks = breaks\_values, labels = abs(breaks\_values)) + theme\_classic() + facet\_grid(~ARM) + geom\_hline(yintercept = 0) + theme(legend.position = "none")

Use ggplot to create the divergent stacked bar chart

RGB colour codes (in order)

c("#145B2F","#57AF62", "#C2A5CE", "#8E4B9B", "#582864", "#2D1433")

### CONCLUSIONS

Data visualisation of available PRO data and completion rate together with clinical events in a patients' journey enables clearer quantification of the potential impact of death and progression on the availability of PRO data, as well as the amount of missing data. This improves the ability to interpret PRO data over time in context of the defined PRO estimand.



<sup>1</sup>FDA: Submitting Patient-Reported Outcome Data in Cancer Clinical Trials: Technical Specifications Document November 2023: https://www.fda.gov/media/173581/download

<sup>2</sup>Coens C, et al; Setting International Standards in Analyzing Patient-Reported Outcomes and Quality of Life Endpoints Data Consortium. International standards for the analysis of quality-of-life and patient-reported outcome endpoints in cancer randomised controlled trials: recommendations of the SISAQOL Consortium. Lancet Oncol. 2020 Feb;21(2):e83-e96.

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