



ARS for Test-Driven Development

Presented by Yann Féat, Statistician, mainanalytics GmbH



Meet the Speaker

Yann Féat

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Yann Féat is a dedicated statistician with over five years of experience in statistics, software engineering, and machine learning, including more than two years in clinical research at mainanalytics GmbH. Yann's passion for statistical reporting and automation drives him to improve the efficiency and accuracy of clinical trial analyses. He holds a Master's degree in Software Engineering and Applied Mathematics and actively contributes to the Regulatory R Package Repository, R Validation Hub.

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- The views and opinions expressed in this presentation are those of the author(s) and do not necessarily reflect the official policy or position of CDISC.
- The author(s) have no real or apparent conflicts of interest to report.





Introduction

Test-driven development (TDD)

development methodology where tests are written before the main code to ensure that all requirements are met



Analysis results standard (ARS)

data model to combine results of statistical analyses with their metadata in a machinereadable way

Combining TDD and ARS into a comprehensive workflow to ensure better consistency, accuracy and efficiency in static reports and in interactive apps



Agenda

- 1. Context and Background
- 2. Introduction to ARS for TDD and their Benefits
- 3. Specification and Testing of ARDs
- 4. Examples
- 5. Conclusion and Q&A

Context and Background

Challenges and innovations in clinical trial reporting

Definitions: business logic vs. service logic



Business logic: handles core computation, including data processing, rule enforcement, and (here) generating ARS-compliant datasets.



Service logic: transforms and formats the outputs from the business layer into TFLs. In the case of interactive apps, also orchestrates interactivity for end-users.





#ClearDataClearImpact

Zafar Mohammad Sami, A. & Féat, Y. (2024). R Shiny from Code to Cloud.

1.amazonaws.com/Archive/2024/Connect/EU/Strasbourg/PAP OS06.pdf

Form modules

PHUSE EU Connect. https://phuse.s3.eu-central-

Main app

Statistical modules

dsfilter module

Current state and challenges of TFL production and QC in the context of clinical trial reporting

TFL production

- Custom script development
- Manual formatting & template adjustment

QC phase

- Manual cross-verification against analysis plans
- Ad-hoc QC & review tools
- Individual interpretation of requirements

Limited standardization & change tracking

- Variability according to individual expertise
- Custom, non-standarized data consolidation

Iterative re-checks & revisions

- Inconsistency in checks
- Extended review cycles
- Fragmented documentation
- Limited traceability



Current state and challenges of TFL production and QC in the context of clinical trial reporting

TFL production

- Custom script dev
- Manual formatting

Critical limitations for interactive visualization and diverse data integration

ation against

tools on of requirements

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Introduction to ARS for TDD and their Benefits

Empowering consistency: the TDD and ARS advantage

Streamlined development process Ensuring that all requirements are met

Early QC

Why TDD with ARS?

Enhanced collaboration through clear specifications for testing and production Traceability and auditability



Unified TDD and ARS workflow

Define requirements & automated tests

- Translate into functionalities using TDD
- Use ARS as blueprint (ensuring clear, interoperable, and compliant outputs)

Develop business logic

- Create ARD*-generating functions
 - Focus on passing all tests
 - Use ADaM datasets as input

Continuous integration & automated QC

- Automated testing and change tracking
- Robust audit trails ensuring high traceability

Apply service logic

- Use validated ARDs to make TFLs
- Integrate in CSR or in interactive app

* Analysis Result Dataset (tabular representation of operation results in the ARS model)



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Specification and Testing of ARDs

Ensuring accuracy through pragmatic testing of ARDs

Prerequisite: specification of analysis methods

CDISC - eTFL Portal

```
"name": "Summary by group of a continuous variable",
"id": "Mth02 ContVar Summ ByGrp",
"operations": [
    "name": "Count of non-missing values",
    "id": "Mth02 ContVar Summ ByGrp 1 n",
    "label": "n",
    "resultPattern": "XX"
    "name": "Mean",
    "id": "Mth02 ContVar Summ ByGrp 2 Mean",
    "label": "Mean",
    "resultPattern": "XX.X"
    "name": "Standard deviation",
    "id": "Mth02 ContVar Summ ByGrp 3 SD",
    "label": "SD",
    "resultPattern": "(XX.XX)"
```

ARS-VS-T01 Summary of Observed and Change from Baseline by Scheduled Visits - Vital Signs Safety Population Parameter (Units) Xanomeline High Dose Xanomeline Low Dose Placebo Visit (N=XX) (N=XX) (N=XX) Parameter (Unit) Baseline n XX XX XX Mean (SD) XX.X (XX.XX) XXX (XXXX) XX.X (XX.XX) Median XX.X XX.X XX.X Min. Max XX.X. XX.X XX.X. XX.X XX.X. XX.X <Visit n> n XX XX XX Mean (SD) XX.X (XX.XX) XX.X (XX.XX) XX.X (XX.XX) Median XX.X XX.X XX.X Min, Max XX.X, XX.X XX.X. XX.X XX.X. XX.X <Visit n Change from Baseline> XX XX XX n Mean (SD) XX.X (XX.XX) XX.X (XX.XX) XXX (XXXX) Median XX.X XX.X XX.X Min. Max XX.X. XX.X XX.X. XX.X XX.X. XX.X

Generated using TFL Designer (Community, v1.0)

"standalone specification of a set of 1 or more statistical operations that can be applied to any analysis variable to generate a set of analysis results"



.....

Page x of y

Definition of ARDs on the CDISC Wiki

Pages / ... / OperationResult

Analysis Result Dataset (ARD)

Created by Richard Marshall, last modified by Lorraine Sobson on Apr 18, 2024

When instances of the OperationResult class are represented in a tabular format, the resultant table may be referred to as an ARD if the following conditions are met:

- There is 1 row for each instance of the OperationResult class (i.e., 1 row per result value).
- Each row includes all the information needed to give context to the result value represented on the row, including at least:
 - the value of the **id** attribute of the analysis that generated and contains the result;
 - the value of the operationId attribute of the instance of the OperationResult class that is represented on the row; and
 - the values of all of the **groupingId**, **groupId**, or **groupValue** attributes that are present in **resultGroups** attribute of the instance of the OperationResult class that is represented on the row.

To aid readability, an ARD may include additional columns containing descriptive information that has been retrieved from instances of other classes using any of the identifier values included on each row. An ARD may contain results from 1 or more analyses. Each of the tables shown in the examples in the OperationResult section is an example of an ARD.





The OperationResult class





Types of unit tests for ARDs

Test type	Example	Purpose		
Structure	Resulting ARD includes required sections: header, body, footer	Confirm dataset adheres to the expected schema		
Column & format	Contains key parameters (e.g., systolic BP, heart rate) in correct formats (e.g., number of decimal places)	Ensure all mandatory fields and formats are present		
Calculation	The mean heart rate in an ARD generated using the 2025-04-04 ADaM snapshot matches the pre-calculated (using the same snapshot) value of 78 BPM.	Validate the correctness of computation results		
Metadata compliance	Relevant metadata fields (e.g., data source) are correctly used.	Guarantee traceability and consistency		
Consistency	Current ARD values are consistent with a fixed earlier snapshot, allowing for natural variation (e.g., the average HR at the current snapshot is within a 10% range from the reference 2025-04-04 ARD snapshot).	Detect regressions and unwanted changes in the data		





Examples

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Example of individual unit test on an ARD

```
# Use the ADVS dataset from an early ADaM snapshot for reproducibility
    advs <- adam_snapshots[[ref_snapshot]]$advs</pre>
 2
    # Run the ARD-generating function to be tested
 3
    ard mean hr <- ard vital summary(advs)Mean Heart Rate[1]
 4
    # Pre-calculated mean heart rate stored in a version-controlled file
 5
    expected mean hr <- test values [[ref snapshot]]$mean heart rate
 6
    # Verify that the ARD mean heart rate is as expected within tolerance
 8
 9 -
    test that("ARD mean heart rate matches expected mean within tolerance", {
      expect true(abs(ard mean - adam mean) <= tolerance,</pre>
10
11
        info = sprintf(paste("ARD Mean Heart Rate = %.2f vs. expected %.2f exceeds",
          "the tolerance of %.4f"), ard_mean_hr, expected_mean_hr, tolerance)
12
13
14 ^ })
```



Pre-calculated test ARD

One ARD per analysis object, one unit test per row

operation Name	resultGroup1 _groupld	resultGroup1 _groupLabel	resultGroup2 _groupId	resultGroup2 _groupLabel	resultGroup3 _groupId	resultGroup3 _groupLabel	rawValue	formatted Value
Count of non-missing values	AnlsGrouping_01 _Trt_3	Xanomeline High Dose	AnlsGrouping_08 _Param_1	Systolic Blood Pressure (mmHg)	AnlsGrouping_09 _Visit_11	End of Treatment	218	218
Mean	AnlsGrouping_01 _Trt_1	Placebo	AnlsGrouping_08 _Param_4	Temperature (C)	AnlsGrouping_09 _Visit_02	Week 2	-0.02454	-0.02
Median	AnlsGrouping_01 _Trt_2	Xanomeline Low Dose	AnlsGrouping_08 _Param_2	Diastolic Blood Pressure (mmHg)	AnlsGrouping_09 _Visit_08	Week 20	-2	-2.0

Central document(s) for TDD

Such a test ARD with pre-calculated values facilitates all kinds of unit tests in an automated manner. It is self-documented, can be easily tracked, and enhances collaboration. It can be implemented using a different software from the main analysis (e.g., SAS vs. R).





Continuous integration

Sample output of unit tests suite run in a Cloud environment





Conclusion and Q&A

Efficiency: standardization, separation of concerns (business vs. service), automation, reusable data

Collaboration: enhanced teamwork, prospective planning, single source of truth Traceability: linking to protocol/SAP, consistency, transparency, standardized metadata





Thank You!

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