
**Data Recipient Report for the
Janssen Clinical Trial Data Set
17335630gal-int-1**

“A Study of the Safety and Effectiveness of Two Doses of Galantamine Versus Placebo in the
Treatment of Patients With Alzheimer’s Disease”

Product Name	Reminyl/Razadyne
Active Substance	Galantamine hydrobromide
Dataset Type	Other
Other Study Identifier	GAL-INT-1
Study Code	17335630gal-int-1
NCT Number	NCT00253188
Reporting Effort	Final
Version	2.0
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1 Introduction

The purpose of this project was to perform anonymization of the Janssen 17335630gal-int-1 clinical trial data set.

The anonymization of this data set was performed to allow the data to be shared with external research teams. Access to clinical trial data provides opportunities to conduct further research that can help advance medical science and improve patient care. This helps ensure the data provided by study participants are used to maximum effect in the creation of knowledge and improving patient care. The data release is subject to certain criteria being met, including a requirement to effectively anonymize the data.

Statistical anonymization was used to preserve the utility required by recipients, while accounting for the context of the data sharing scenario [2]. Unlike a rules-based framework that removes dates (except years) and aggregates all ages over 89 as 90 or older, such as HIPAA Safe Harbor, this approach is adaptive to population distributions, sample size, and the desired utility of the anonymized data.

The data sharing environment and contracts in place with the data recipient are assumed to be at a level which would result in a Privacy and Security Context Assessment score of High and a Recipient Trust Context Assessment score of Medium.

This report describes the anonymization approach used for the study 17335630gal-int-1, based on the re-identification risk determination that was performed on the data.

1.1 Data Set Model

The data set described in this report for study 17335630gal-int-1 was received in a custom raw data format.

1.2 Definitions

Definitions of key terms (such as the different types of identifiers) and acronyms are provided in Section A *Definitions*. Additional terms and definitions are provided elsewhere [1].

2 Anonymization Process

2.1 Use of Software

The analysis described in this report was performed using a re-identification risk measurement software application.

2.2 Supporting Documentation

The following documents were provided to assist with the analysis:

- 17335630gal-int-1 Transformation Summary
- Annotated CRF

2.3 Output Format of Anonymized Datasets

All dataset anonymization was performed within the SAS (Statistical Analysis System) native data file format (extension “.sas7bdat”). Datasets received in SAS version 5 (V5) or version 8 (V8) transport file format (extension “.xpt”) must first be converted to .sas7bdat for processing. Following de-identification, all datasets are converted from .sas7bdat to .xpt for delivery. For datasets originally received in .xpt format, this conversion should not pose a problem. However, for datasets received in non-xpt format, inherent limitations in the .xpt format may require modifications.

Based on the definition of the format, conversion of a dataset to XPT transport file format may require modification of the following in the anonymized datasets:

1. Shortening the dataset names,
2. Shortening variable names in the datasets,
3. Shortening dataset or variable labels,
4. Splitting long character values into new variables.

2.4 Transformations

In order to bring the risk of re-identification below the determined threshold, some transformations were required on the dataset. The transformations are described based on the indirect identifiers used in the risk measurement. In all cases, modifications to these indirect identifiers are applied to all other linked fields, e.g. where country is suppressed, fields containing brand- or region-specific drug names will also be suppressed as they are linked to geography.

The anonymization strategy required the following modifications to the original datasets:

Identifier	Transformation
Subject IDs (CRFID)	Masked
Free-text	Suppressed
Patient dates	PHUSE shifted
Date of Birth	PHUSE shifted and Generalized to 5 year intervals
Race	Suppressed
Accommodation	Suppressed
Education	Suppressed
Caregiver Date of Birth	Suppressed
Caregiver gender	Suppressed
Caregiver Relationship	Suppressed

2.5 Implemented Transformation Types

The following data transformations have been applied in this dataset:

Masking Masking of the unique subject ID was performed using Format-Preserving Encryption (FPE). This type of encryption creates an encrypted value that has the same length as the original ID.

Generalization Reduce the precision of a field.

For this specific project, the year of birth of subjects was first PHUSE date shifted then generalized to 5-year intervals. Table 1 summarizes the mapping of year of birth to generalized year of birth for values greater or equal to 1900.

PHUSE date shifting Offset a date value according to the scheme defined in the Pharmaceutical Users Software Exchange (PHUSE) CDISC SDTM anonymization standard [3]. This scheme determines a delta for each patient based on a difference between a date in the trial available for all patients (in this case the first visit date) and an anchor date (in this case, 15 Feb 1997).

Suppression The original value is replaced with an empty cell. The following type of suppression was applied for this project:

global suppression (GS): Occurs when risk measurement determines that no suitable generalized value can be retained and all values in the column are therefore suppressed.

Please see the file “17335630gal-int-1 Transformation Summary.csv” for a catalog of all transformations applied to the dataset.

Year of Birth	Generalized Year of Birth (BIRTH_D)
1900 \leq YOB < 1905	1900
1905 \leq YOB < 1910	1905
1910 \leq YOB < 1915	1910
1915 \leq YOB < 1920	1915
1920 \leq YOB < 1925	1920
1925 \leq YOB < 1930	1925
1930 \leq YOB < 1935	1930
1935 \leq YOB < 1940	1935
1940 \leq YOB < 1945	1940
1945 \leq YOB < 1950	1945
1950 \leq YOB < 1955	1950
1955 \leq YOB < 1960	1955

Table 1: Year of birth generalization.

3 Conclusions

The re-identification risk of the Janssen 17335630gal-int-1 clinical trial database, after the anonymization as described in this report, is below the data risk threshold given the assumed level of mitigating controls and motives and capacity in the context of the data sharing environment.

References

- [1] Khaled El Emam. *Guide to the De-Identification of Personal Health Information*. CRC Press (Auerbach), 2013.
- [2] International Standards Organization. ISO/IEC 27559:2022: Information security, cybersecurity and privacy protection – Privacy enhancing data de-identification framework. Technical report, ISO, 2022.
- [3] PhUSE De-Identification Working Group. De-Identification Standards for CDISC SDTM 3.2. Technical report, 2015.
- [4] Pierangela Samarati. Protecting Respondents' Identities in Microdata Release. *IEEE Transactions on Knowledge and Data Engineering*, 13(6):1010–1027, 2001.
- [5] L. Sweeney. k-Anonymity: A Model for Protecting Privacy. *International Journal on Uncertainty, Fuzziness and Knowledge-based Systems*, 10(5):557–570, 2002.

A Definitions

A.1 Acronyms

FPE Format-Preserving Encryption

PHUSE Pharmaceutical Users Software Exchange

A.2 Identifiers

It is useful to differentiate among the different types of variables in a disclosed data set or document. The way the variables are handled during the risk measurement and anonymization process will depend on how they are categorized.

A distinction is made among three types of variables [4, 5]:

Directly identifying variables. One or more direct identifiers can be used to uniquely identify an individual, either by themselves or in combination with other readily available information. In clinical trial data sets and documents, the only patient direct identifier will likely be the subject ID. There will be direct identifiers pertaining to staff and investigators; however, these are treated differently than patient information.

Indirectly identifying variables. The indirect identifiers are attributes that, together with other attributes that can be in the dataset or external to it, enable unique identification of a data subject within a specific operational context.

Examples of indirect identifiers include sex, date of birth or age, locations (such as postal codes, census geography, information about proximity to known or unique landmarks), language spoken at home, ethnic origin, aboriginal identity, total years of schooling, marital status, criminal history, total income, visible minority status, event dates (such as admission, discharge, procedure, death, specimen collection, visit/encounter), codes (such as diagnosis codes, procedure codes, and adverse event codes), country of birth, birth weight, and birth plurality.

Other variables. These are the variables that are not really useful for determining an individual's identity. They may be clinically relevant or not.

A.3 Glossary

data recipient The data recipient is the researcher who accesses the anonymized data to perform an analysis.

Privacy and Security Context Assessment A questionnaire that evaluates the privacy and security controls in place for a data recipient.

Recipient Trust Context Assessment A questionnaire that evaluates the motives, capacity, and contracts in place with regard to data recipient performing a re-identification attack.

B Datasets Delivered in 17335630gal-int-1

Dataset	Number of Rows
ADAS	345721
ADMMED	2433
AE	2925
CARCONS	1540
CARETIME	33379
CMRI	4359
COTHER	3890
DAD	86802
DEATH	6
DEPENDS	8541
DEVIATN	297
DIAGNOS	742
DISEASES	10241
ECG	2587
ECGABN	152
ECGPAP	11730
EMERROOM	89
GENOTYPE	563
GWBS	39446
HABIT	752
HOSPITAL	141
INEX	13518
INVEST	149
LABCOM	2531
LABNOR	143
LABREF	4994
LABRES	157794

Dataset	Number of Rows
LABSAM	5657
LABURI	27129
MMSE	9012
NEUREXAM	6751
NURSHOME	174
PAIDHOME	4146
PHONE	1886
PHYSEXAM	31437
PKMISSED	562
PLAADM	2241
PLARES	0
PLASAM	3366
PROFVIS	9518
RELAE	246
RELATED	657
REMARK	2274
RESQUEST	4161
SUBJCHAR	753
TRLTERM	881
VISIT	6840
VITSIGN	4929
XSUB	653

Table 2: List tables considered and the number of rows in each.