Identifying Predictive Factors of Patient Dropout in Alzheimer's Disease Clinical Trials

Patricia J Allen, MS¹, Iman Abba¹, Corinne Ahlberg, MS¹, Luke Benz¹, Hiuyan Lau, PhD¹, Jingshu Liu¹, Fareed Melhem¹, Nahome Fisseha, PharmD², Hana Florian, MD² ¹Acorn AI by Medidata, a Dassault Systèmes Company, New York, NY, USA; ²AbbVie Inc., North Chicago, IL, USA;

Study population

- 8103 patients from 665 clinical trial sites in 7 completed, phase 3, interventional (oral) AD trials > were included in the analysis (Table 1)
- The treatment period across trials was 1.4 ± 1.6 years (mean \pm SD) and the dropout rate was **11.2 ± 10.8%**

Table 1. Patient and Site Characteristics

| Patients | Dropout (N = 1982) | Complete (N = 6121) |
|------------------------|-----------------------|------------------------|
| Age (mean ± SD) | 74.5 ± 7.1 | 73.2 ± 7.5 |
| Sex: Female | 1133 (57.2%) | 3598 (58.8%) |
| Race | | |
| White | 1817 (01 7%) | 5506 (01 1%) |
| White | 1817 (91.7%) | 5596 (91.4%) |
| Asian | 57 (2.9%) | 240 (3.9%) |
| Black | 62 (3.1%) | 101 (1.7%) |
| Missing/Other | 46 (2.3%) | 184 (3.0%) |
| Education | 14.0 ± 3.5 | 13.8 ± 3.8 |
| Year of Enrollment | | |
| | | |
| 2010-2014 | 1187 (59.9%) | 2824 (46.1%) |
| 2015-2018 | 795 (40.1%) | 3297 (53.9%) |
| MMSE (mean ± SD) | 24.7 ± 5.8 | 25.0 ± 5.2 |

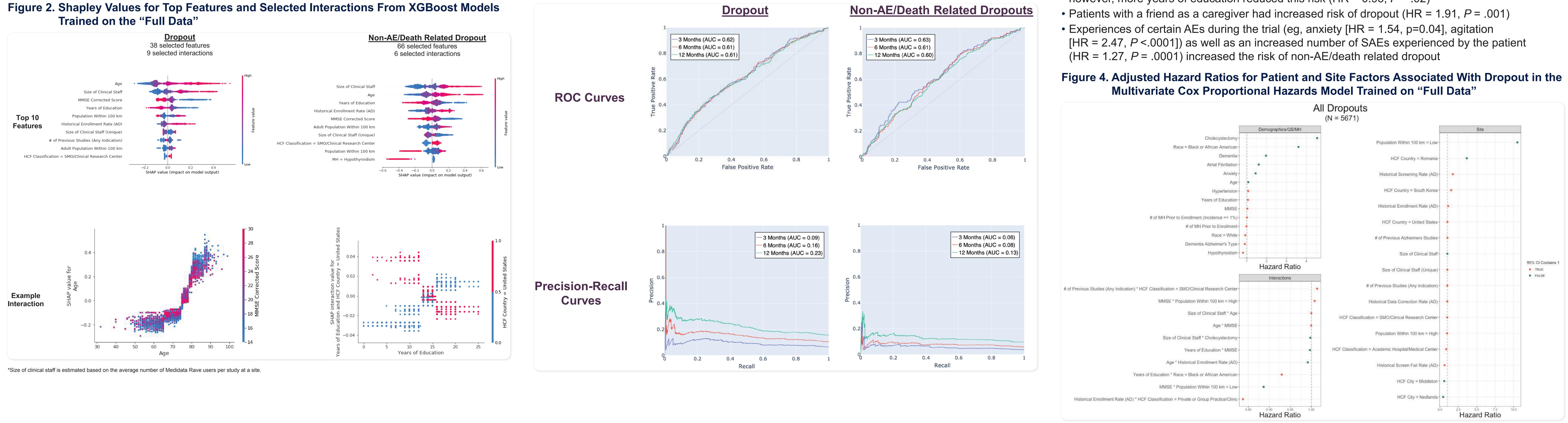
Z Background

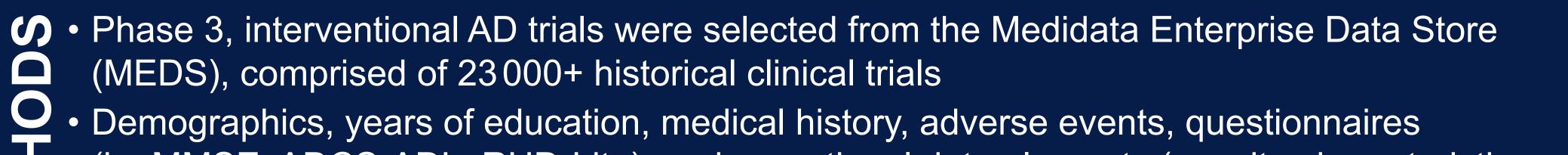
- Patient dropout is a common challenge in Alzheimer's Disease (AD) clinical trials that can lead to trial delays, increased costs, and potentially biased trial results¹
- Previous studies have shown that certain patient and site characteristics influence the risk \sim of patient dropout in AD trials, such as age^{2,3}, race^{2,4}, education^{2,4}, site facility type⁴, and caregiver relationship⁵
- The objectives of this analysis were to use pooled clinical trial data to evaluate how well dropout risk can be predicted at the time of randomization and to identify additional predictive factors of patient dropout in AD trials
- Predictive models and an understanding of dropout risk factors can be used to improve operational aspects of future trial designs as well as to support patients with retention strategies during future trials

Feature Selection

- Size of clinical staff*, patient age, and years of education were among the most important features The models were observed to predict dropouts at an acceptable level above random guessing identified for prediction of dropout, and specifically, non-AE/death related dropout with an ROC-AUC \geq 0.60 within 3, 6, and 12 months
- The risk associated with low MMSE score at baseline varied with age where older patients (>80 years) with low baseline MMSE scores had a higher risk of dropout
- Fewer years of education (≤15 years) for patients in the US was associated with a higher risk of non-AE/death related dropout

Trained on the "Full Data"





- (ie, MMSE, ADCS-ADL, RUD-Lite), and operational data elements (eg, site characteristics and historical performance) were standardized across trials using proprietary machine S learning algorithms combined with human review
- Patients were randomly split into a 70/30 train-test split for the analysis. Equal representation of studies and dropout rates were ensured between the train and test sets
- Several multivariate Cox proportional hazards models were explored to predict the risk of dropout within 3, 6, and 12 months of randomization and to evaluate the effects of patient and site level factors on dropout (Figure 1)
- Stratification and non-linear transformations were used respectively to correct categorical and continuous variables that violated the proportional hazards assumption

Model Performance

• However, the probability of dropout was typically overestimated, leading to more false positives and reduced precision overall (PR-AUC < 0.24)

Figure 3. ROC and PR Curves for the Multivariate Cox Proportional Hazards Models Trained on "Full Data"

| Objectives | Model | Data | Outcomes |
|---|---|--|----------------------|
| 1. Feature selection and identification of interactions | XGBoost w/Cox Proportional Hazards objective function | 8103 patients w/MMSE scores "Full Data" | Dropout |
| | | | Non-AE/Death Dropout |
| | | 4630 patients w/MMSE + ADCS-ADL + RUD-Lite scores "QS Data" | Dropout |
| | | | Non-AE/Death Dropout |
| 2. Assess predictive power and impact of baseline features | Cox Proportional Hazards Regression | 8103 patients w/MMSE scores "Full Data" | Dropout |
| | | | Non-AE/Death Dropout |
| | | 4630 patients w/MMSE + ADCS-ADL + RUD-Lite scores "QS Data" | Dropout |
| | | | Non-AE/Death Dropout |
| 3. Assess impact of time-varying features | Time-variant Cox Proportional Hazards Regression | 8103 patients w/MMSE scores | Dropout |
| | | "Full Data" | Non-AE/Death Dropout |
| | | 4630 patients w/MMSE + ADCS-ADL + RUD-Lite scores "QS Data" | Dropout |
| | | | Non-AE/Death Dropout |

• A limited set of patient and site features can be used to estimate patient dropout risk within 3, 6, and 12 months of randomization

 Important interactions exist between patient and site factors which should be considered in the evaluation of patient dropout risk

• Previous findings of patient and site factors associated with dropout^{2–5} are reflected in this study, in addition to new factors such as site location characteristics and anxiety before/during the trial Many of these risk factors remain for dropouts that resulted from reasons other than AE/death Further investigation into these and other factors associated with patient dropout is warranted to improve understanding and to further define patient retention strategies in AD trials

Predictive Factors

S

- Patient age (adjusted hazard ratio [HR] 1.07, P = .005), race (HR = 3.60, P = .04), certain medical histories (eg, cholecystectomy [HR = 4.54, P = .004], anxiety [HR = 1.44, P = .006]), low surrounding population density (HR = 10.50, P = .04), and the size of clinical staff (HR = 1.01, P = .02) were among the factors associated with increased risk of dropout at the time of randomization
- Patients in the US had increased risk of non-AE/death related dropout (HR = 3.02, P = .03); however, more years of education reduced this risk (HR = 0.95, P = .02)

Presented at the Alzheimer's Association International Conference, July 26–30, 2021, Denver, Colorado, United States and Online bbVie and Medidata Disclosur

Vie funded the research for this study and participated in the interpretation of data. Medidata's data contributions were analyzed for the research. AbbVie and Medidata, participated in the study of nd writing, reviewing, and approving this abstract for submission. All authors had access to the data; participated in the development, review, and approval of this presentation, and agreed to submit this presentation

PJA, LB, IA, JL, HL, CA, and FM are salaried employees of Acorn AI, by Medidata, a Dassault Systèmes company. **NF** and **HF** are employees of AbbVie and may hold AbbVie stock and/or stock options.

References

- lisease clinical trials. Alzheimers Res The 2010:2(6):34. Published 2010 Dec 21 doi:10.1186/alzrt58
- Dis Assoc Disord. 2019:33(4):299–306. UUI. 10. 1097/VVAD.0000
- 3. Edland SD, Emond JA, Aisen PS, Petersen RC NIA-funded Alzheimer centers are more efficient than commercial clinical recruitment sites for conducting secondary prevention trials of dementia Alzheimer Dis Assoc Disord. 2010;24(2):159–64. doi:10.1097/WAD.0b013e3181c9983f.
- . Bernstein OM. Grill JD. Gillen DL. Recruitment and ntion of participant and study partner dyads in tw nultinational Alzheimer's disease registration trials Alzheimers Res Ther. 2021:13(1):16. Published 202 Jan 8. doi:10.1186/s13195-020-00762-8
- . Colev N. Gardette V. Toulza O. et al. Predictive patients. The REAL.FR study. *Neuroepidemiology* 2008;31(2):69–79. doi:10.1159/000144087



QR code or utilize the following link nload an electronic version of th code expiration: June 26, 2022. ubmit a medical question, please visi bviemedinfo.com

abbvie