



# Modeling Dropout from Longitudinal Adverse Event Data

## Selecting Optimal Titration Regimens

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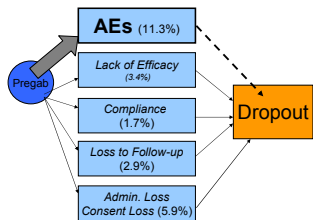
### INTRODUCTION

- Pregabalin offers a new approach in the treatment of Generalized Anxiety Disorder (GAD)<sup>1</sup>.
- Adverse events (AEs) are the predominant component of study withdrawal (dropout) contributing with at least 60% to total dropout incidence for this indication, considerably more than other dropout components/categories<sup>2-6</sup>(Figure 1).
- Pregabalin AE incidence is dose dependent; as AEs have been modeled, we postulate a dropout model that incorporates severity of adverse effects.
- Dose was titrated based on clinical judgment; modeling approach will allow titration schedule optimization.
- Dropout analysis represents analysis of competing risk, we present a component specific approach, as if dropout arises from a specific risk of interest (AEs), the dose-dependent source of risk.

### OBJECTIVE

- Model the dropout across the various titration schemes and evaluate the daily AE scores, a time-dependent covariate, as a predictor in the model.

Fig 1. Generalized Anxiety Disorder (GAD) Dropout Incidence (ITT)



### Pregabalin Dataset-GAD Phase II studies<sup>2-6</sup>

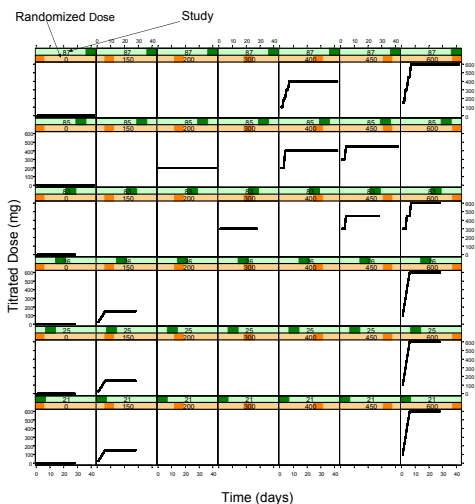
- Five comparator trials with benzodiazepines or SSRI (3 alprazolam, lorazepam and venflaxamine) and a schedule optimization study (TID vs. BID) informed the model.
- 1630 predominantly female, 18 to 64 years old subjects receiving placebo or pregabalin all reported daily subjective AE scores [no (0), mild (1), moderate (2) and severe (3)] across the 4 or 6 week trials.
- Studies employed varying dose within the first week of the protocols (Fig. 2)

### Adverse Events Incidence Self-Reporting

	Placebo	150 mg/day	600 mg/day
Dizziness (%)	6-8	10-23	29-39
Somnolence (%)	11	15-29	36-50
Others* (%)	<20	<20	<20

\*Headache, Dry Mouth, Lack of Coordination, Nausea and Vomiting

Fig 2. Titrated Doses Across Study Arms (Randomized Doses)



### METHODS

#### Discrete Survival and Conditional (Hazard) Probabilities

- Time to subject dropout (T) is modeled as discrete survival variable. S(t) is the probability that an individual survives in the study (does not drop out) beyond time t<sub>j</sub>, where j indexes study day.

$$S(t) = Pr(T > t)$$

- Survival probability is related to hazard h(t<sub>j</sub>), the probability of the event (dropout) occurring in the time interval j, provided it has not occurred prior to j, among those "at-risk" (in study, during interval j). S(t) is also the product of conditional survival probabilities, i.e., sequential survival across each individual period t<sub>j</sub>.

$$h(t) = Pr(T = t | T \geq t) = \frac{Pr(T = t)}{Pr(T \geq t)} = 1 - \frac{S(t)}{S(t-1)} = 1 - e^{-\int_{t-1}^t g(t, AE_{MAX}=m) dt}$$

- Assuming that the probability of dropout (likelihood) is only dependent on the covariate (AE score) in a particular time interval, it is:

$$Pr(T = t) = h(t) \prod_{j=1}^{t-1} (1 - h(t_j))$$

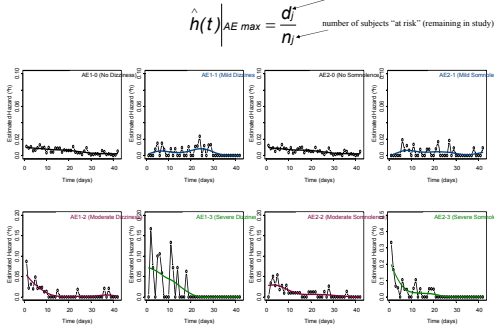
#### Conditional Probability Cumulative Hazard Model

- A discrete-time survival model was fit to GAD dropout time, assuming that the hazard depends on the current self-reported AE score and time since first study day.

$$h(t) = Pr(T = t | T \geq t, AE_{MAX} = m) = 1 - \exp\left(-\int_{t-1}^t g(t, AE_{MAX}=m) dt\right)$$

- T is the time of dropout, AE<sub>MAX</sub> the maximum AE score for the patient at time t<sub>j</sub> and g(t<sub>j</sub>, AE<sub>MAX</sub>=m) the hazard probabilities conditioned on the last observed AE<sub>MAX</sub> score (a parsimonious severity covariate).
- Nonparametric examination of risk (see Figure 3) guided the selection of the (parametric) hazard model

Fig 3. Nonparametric Estimates of the Dropout Risk- Regression of Hazard Probabilities vs. Time Given Dizziness and Somnolence, Time-Dependent AE Covariates



- Linear (exponential) and several time-dependent hazard probability models including the Weibull, loglogistic, linear-exponential and Gompertz were considered.
- Gompertz model (exponentially decreasing hazard with time) where λ is a hazard intercept and γ the slope with time, was selected based on visual inspection of the fitted vs. observed hazard plots, NONMEM OFV, predictive check model simulations and bootstrap confidence intervals. (Figures 4 and 5)

$$g(t_j, AE_{MAX} = m) = \exp(\lambda m + \gamma m \cdot t)$$

- The LAPLACIAN method as implemented in NONMEM V (Icon Corp., Ellicott City, MD) provided the maximum likelihood (-2LL) parameter estimates for the hazard parameters<sup>7-8</sup>.

Fig 4. Fitted Hazard Probability Distribution Models to the Observed Conditional Probabilities

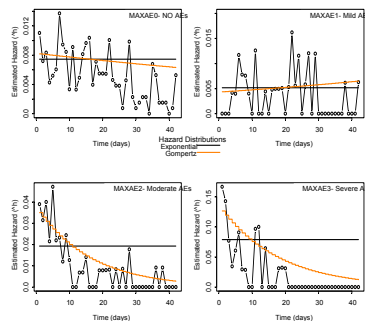
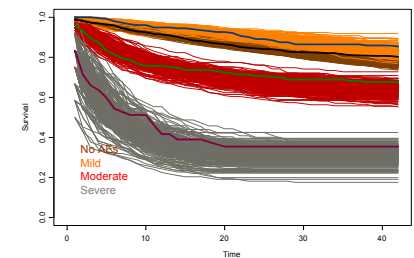


Fig 5. Simulation of Dropout (Gompertz Hazard Model) using Observed MaxAEs- Predictive Check (200 Simulations) with KM Nonparametric Survival Estimates (thick lines)



#### Prospective Simulations

- To generate prospective titration schemes AE data were simulated according to a 2-part incidence and severity AE model<sup>9</sup>. Simulated AEs and resulting dropouts are depicted below.

Fig 6. Dropout from Simulated AEs (3 Part Model) for Placebo, 450-600 mg Dose Groups/Titration Schemes. Blue line depicts the observed Kaplan-Meier dropout estimates.

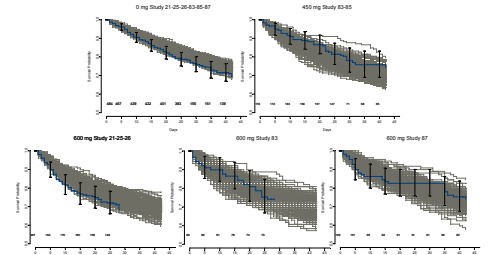
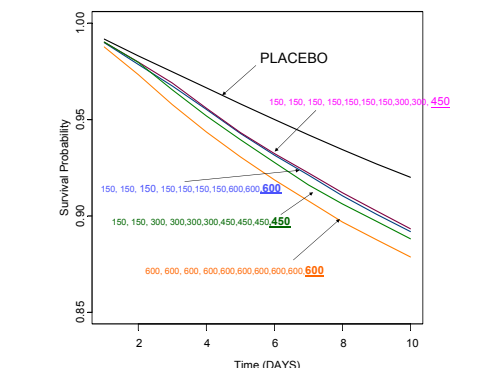


Fig 7. Mean Dropout from Prospective 7-day Titration Schemes (300 Simulations) using the 3-Part AE-Dropout Model



### RESULTS

- Exponentially decreasing dropout risk (hazard) probability model (Gompertz) adequately described the dropout across GAD studies.
- Hazard was highest with high severity of either dizziness and somnolence (AE<sub>MAX</sub>) decreasing with time for individuals with moderate or severe AEs; subjects reporting severe dizziness or somnolence on day 1 exhibit a 21% per day chance of dropout, which decreases to 9% and 1% per day on days 10 and 30 respectively.
- Patients reporting mild or no adverse events were estimated to exhibit a constant but low dropout risk, <1% per day at all times.
- Prospective studies can be planned/optimized in respect to AE and dropout across other indications.

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