

Modelling and Simulation of the incidence of adverse events in clinical trials

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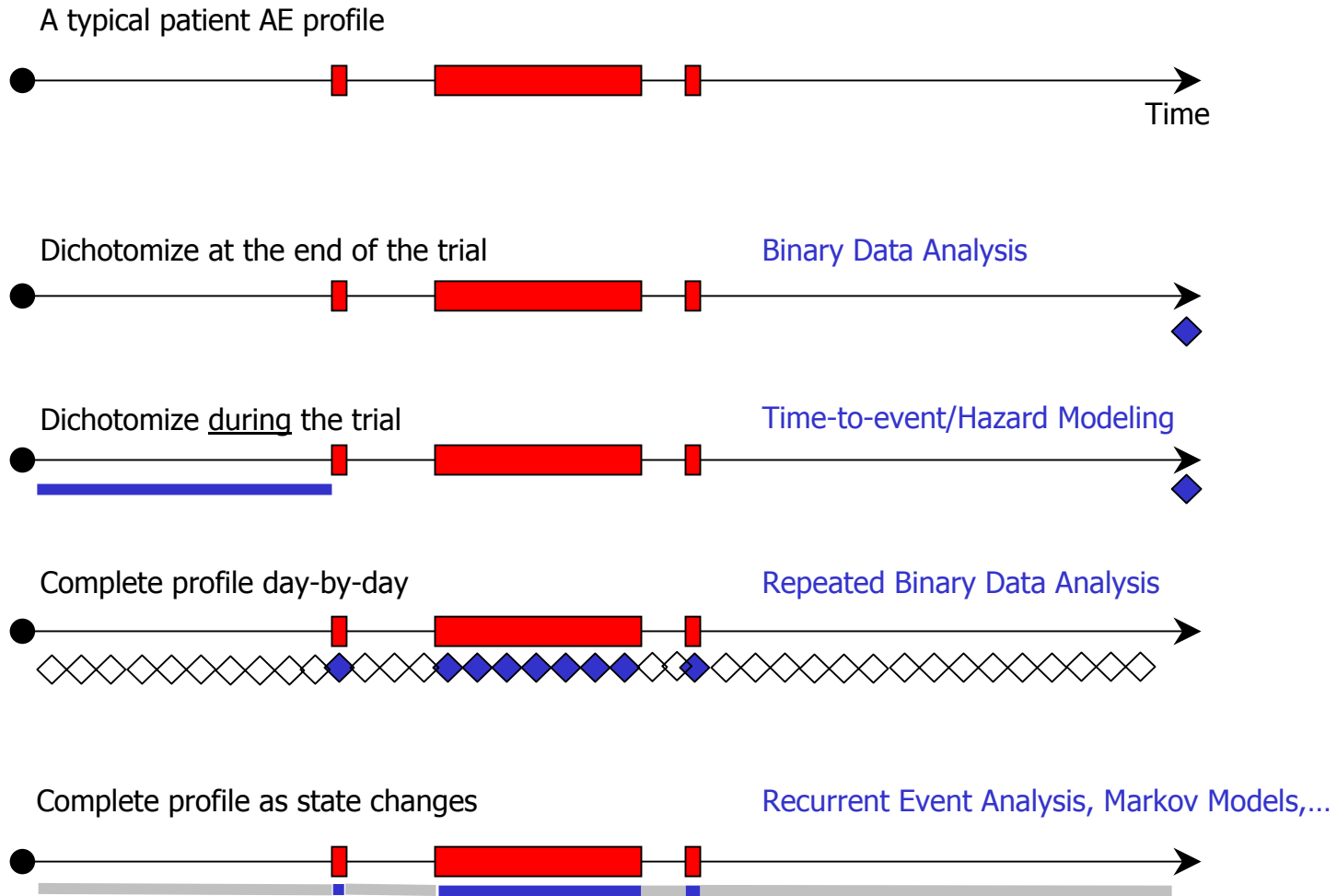
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Introduction

- ◆ Drug Y is an improved version of a marketed drug X
- ◆ A specific adverse event is related to supra-efficacious drug levels and is expected for drug Y as well.
- ◆ In planned clinical trials with drug Y, we want to control the incidence of the adverse event, to balance efficacy and safety
- ◆ PK/PD model using data of drug X (~2500 patients, 10 trials)
- ◆ Predict adverse event in clinical trials with Y (based on a pharmacokinetic and pharmacological rationale)

How to model adverse events?



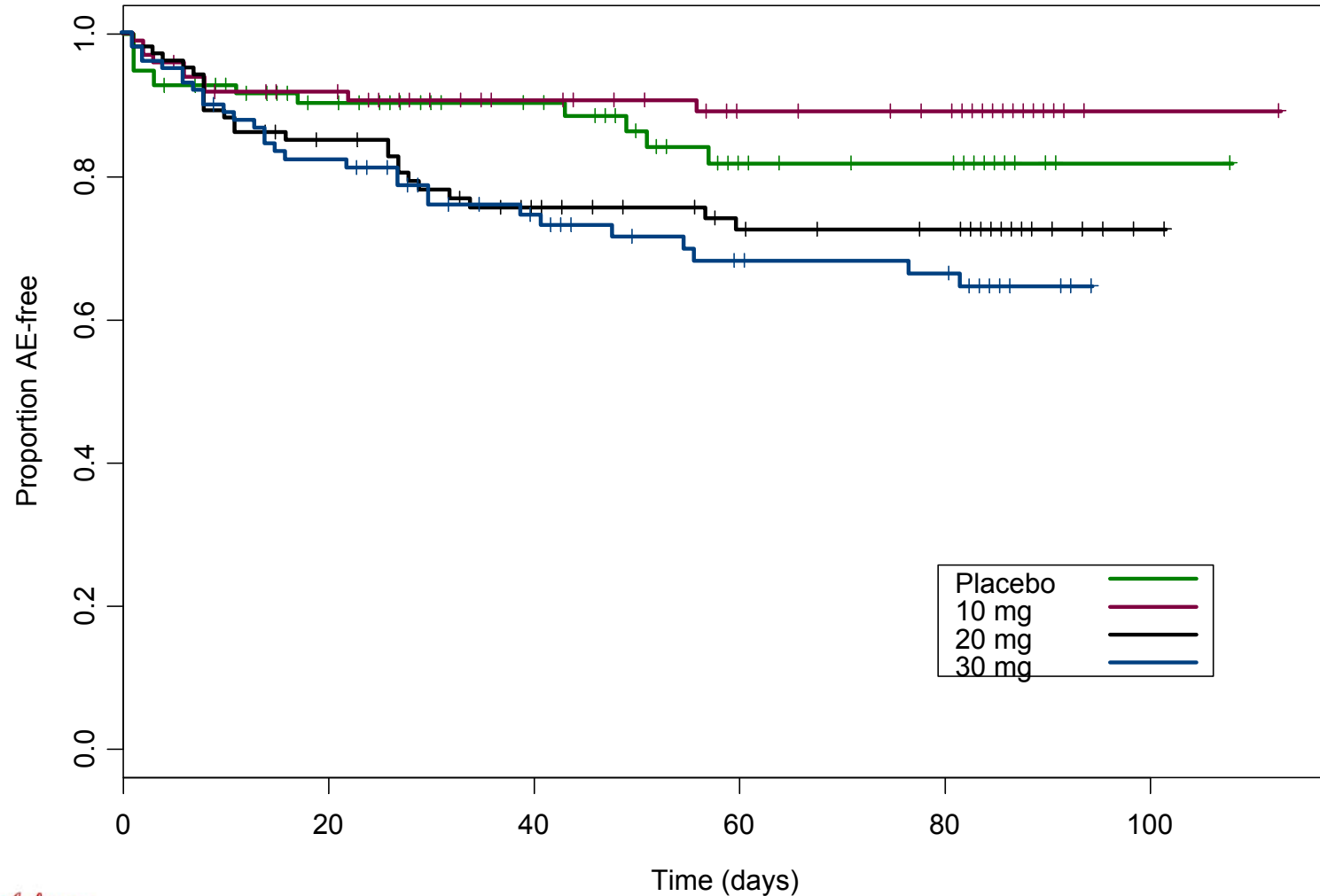
The Data of Drug X

- ◆ Double-blind placebo controlled trial with 3 doses

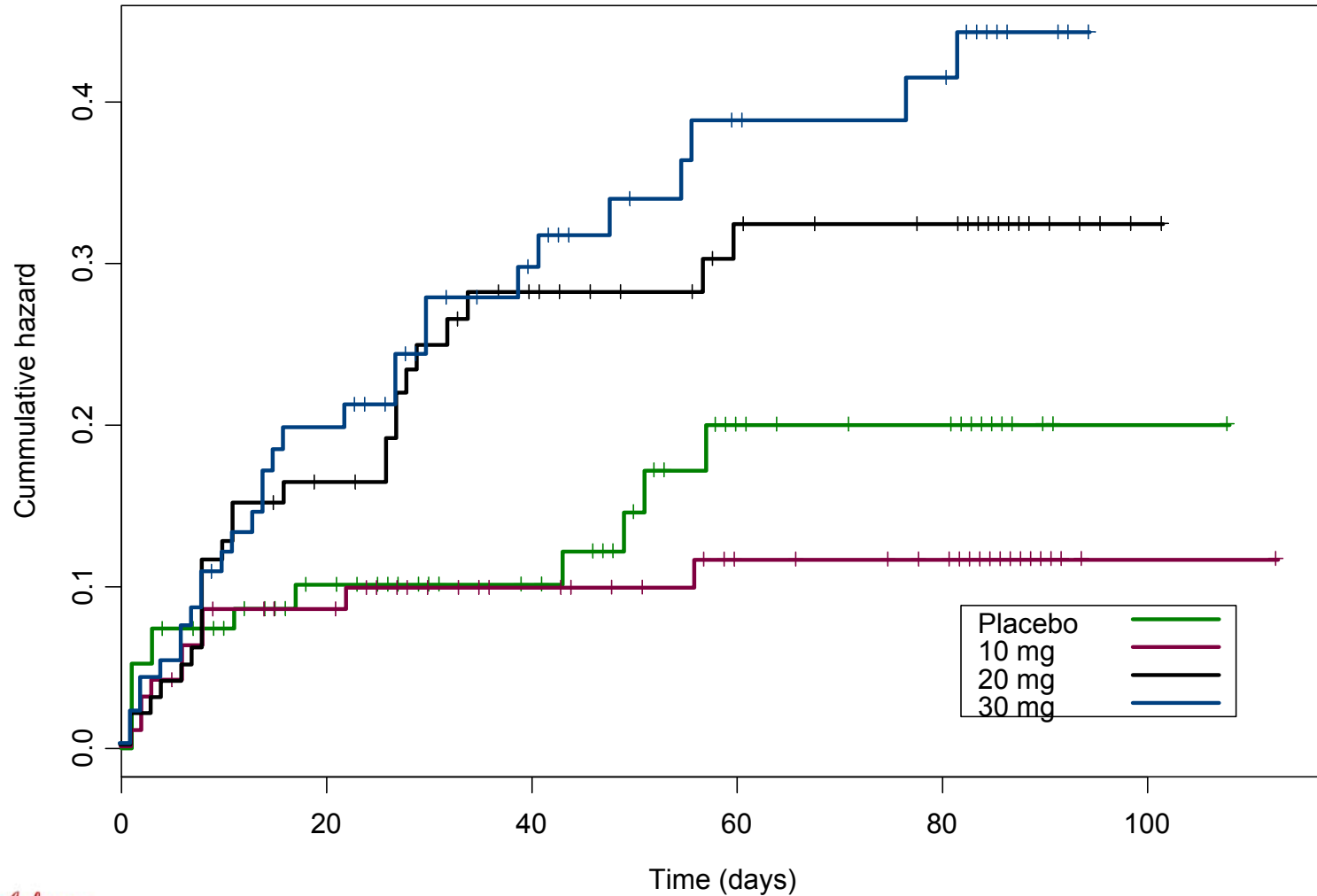
<u>Treatment</u>	<u>N</u>	<u>Drop-out(%)</u>	<u>AE(%)</u>
Placebo	98	67(68)	13(13)
10 mg	99	51(51)	10(10)
20 mg	103	54(51)	25(24)
30 mg	100	52(52)	29(29)

$$AE = \frac{\# \text{ of patients with at least 1 AE-episode}}{\# \text{ of patients randomized}}$$

Kaplan-Meier plot for time to first AE



Cumulative hazard



What measure of drug exposure?

- ◆ No scientific rationale available
- ◆ As the adverse event develops over time, an integrated measure of long term exposure was chosen

Individual - Average Steady-State Concentration (C_{av_i})

$$C_{av_i} = \text{Dose}/Cl_i/\tau$$

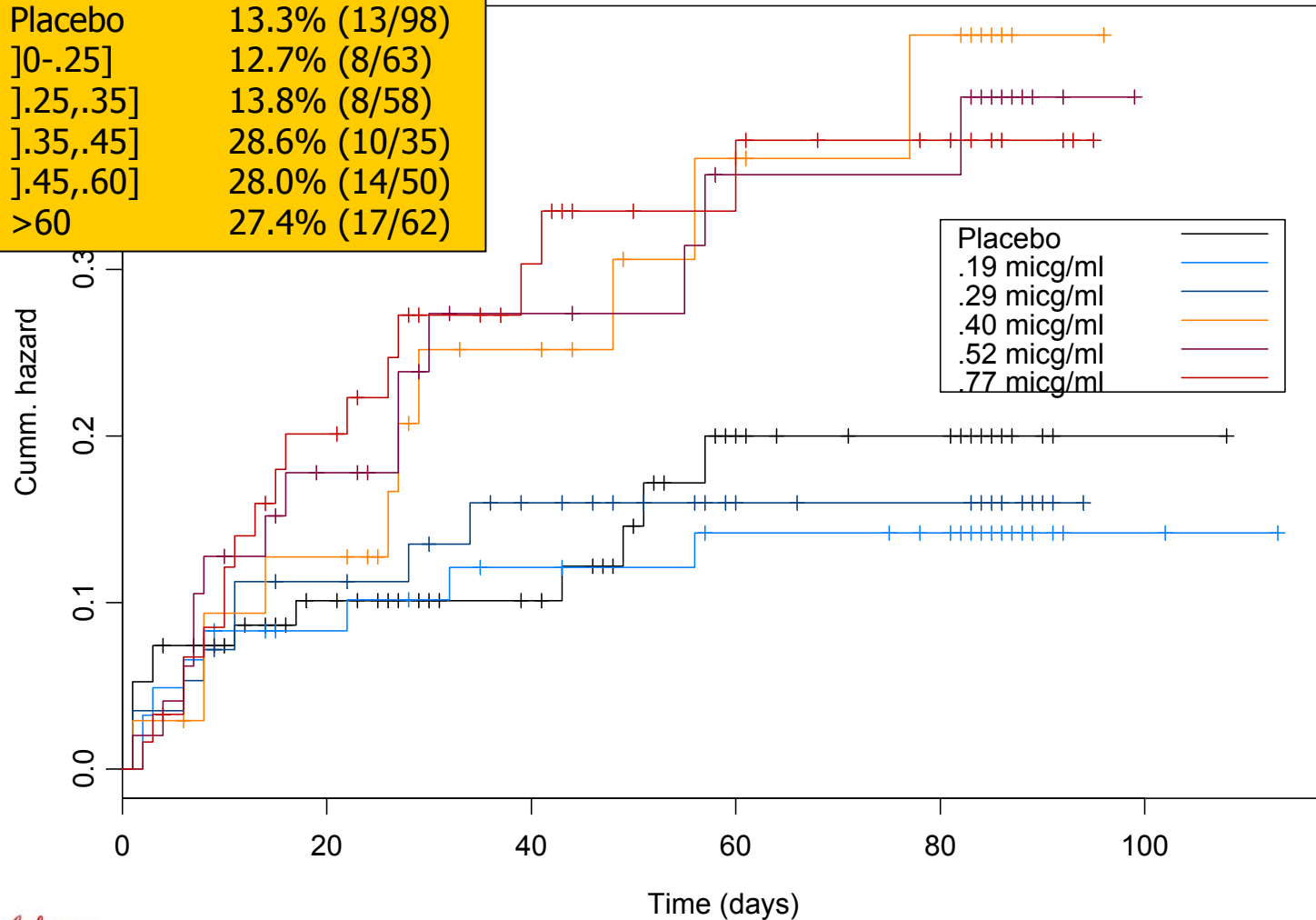
Cl_i : Individual Clearance

τ : Dosing interval

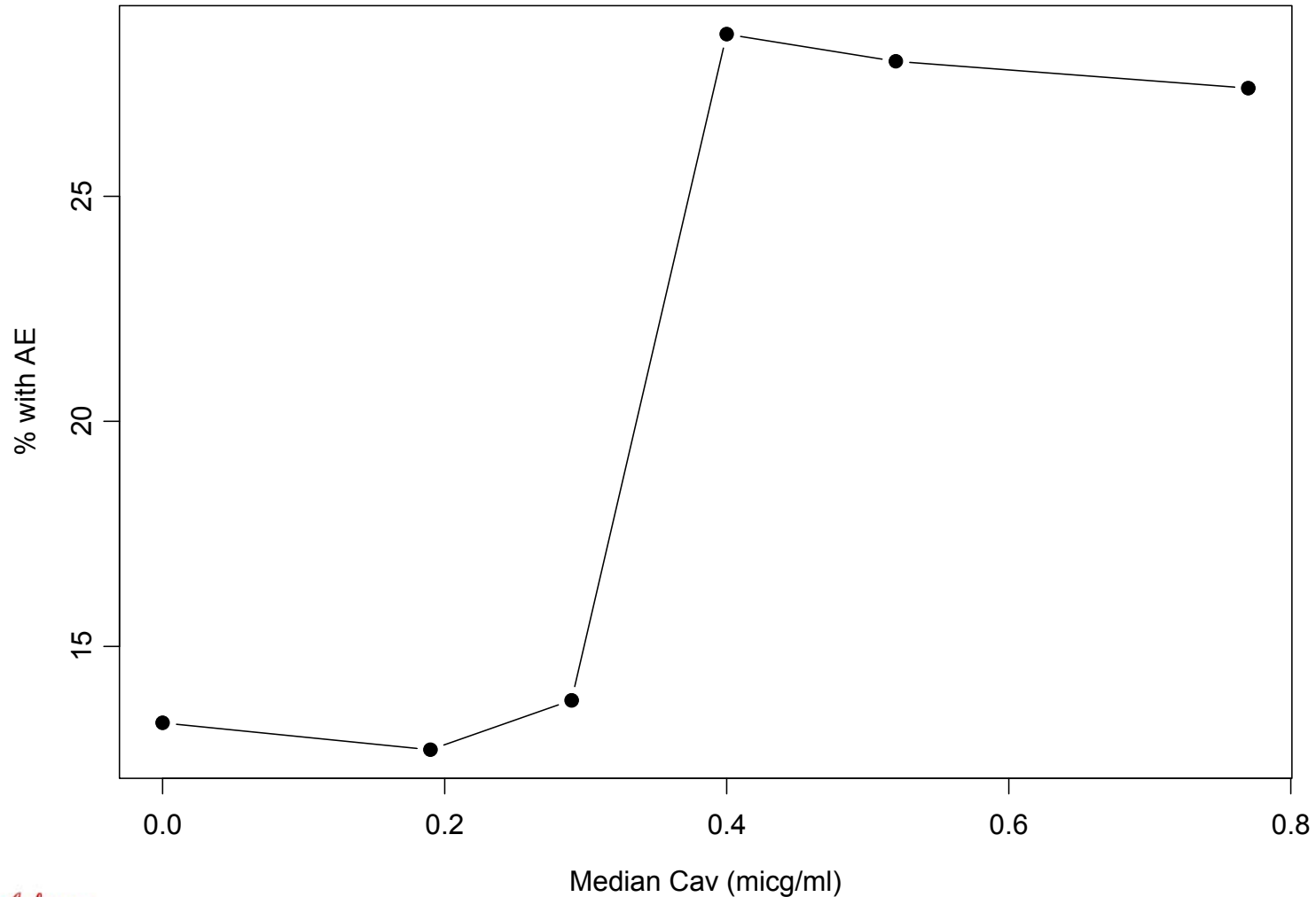
Bayesian Individual Predictions
Population PK model

Cumulative hazard x Exposure

Cav ($\mu\text{g/ml}$)	Incidence
Placebo	13.3% (13/98)
]0-.25]	12.7% (8/63)
] .25,.35]	13.8% (8/58)
] .35,.45]	28.6% (10/35)
] .45,.60]	28.0% (14/50)
>60	27.4% (17/62)



Incidence of AE x Exposure



Modeling the AE risk

$$\log \lambda(t, c_i) = f(t) + g(t, c_i)$$

t : time since trial start

c_i : individual average steady state concentration

Time dependency

$$f(t) = \beta_0 + \beta_1 t$$

Drug effect - logistic

$$g(c_i) = \frac{\theta_1}{1 + \exp\left(\frac{\theta_2 - c_i}{\theta_3}\right)}$$

θ_1 : maximal effect

θ_2 : EC50

θ_3 : scale parameter (steepness)

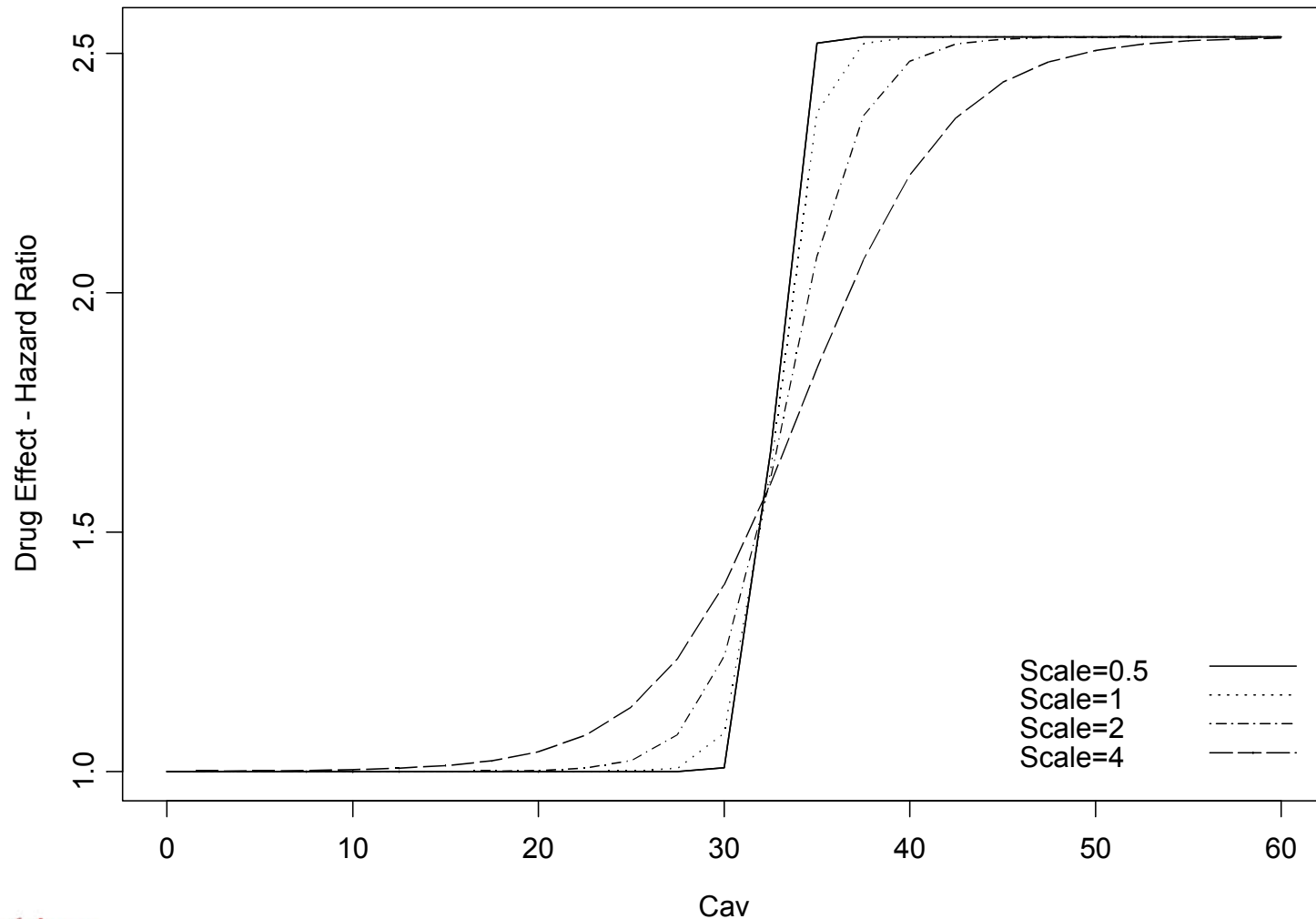
Drug effect - threshold

$$g(c_i) = I(c_i > \theta_1) \cdot \theta_2$$

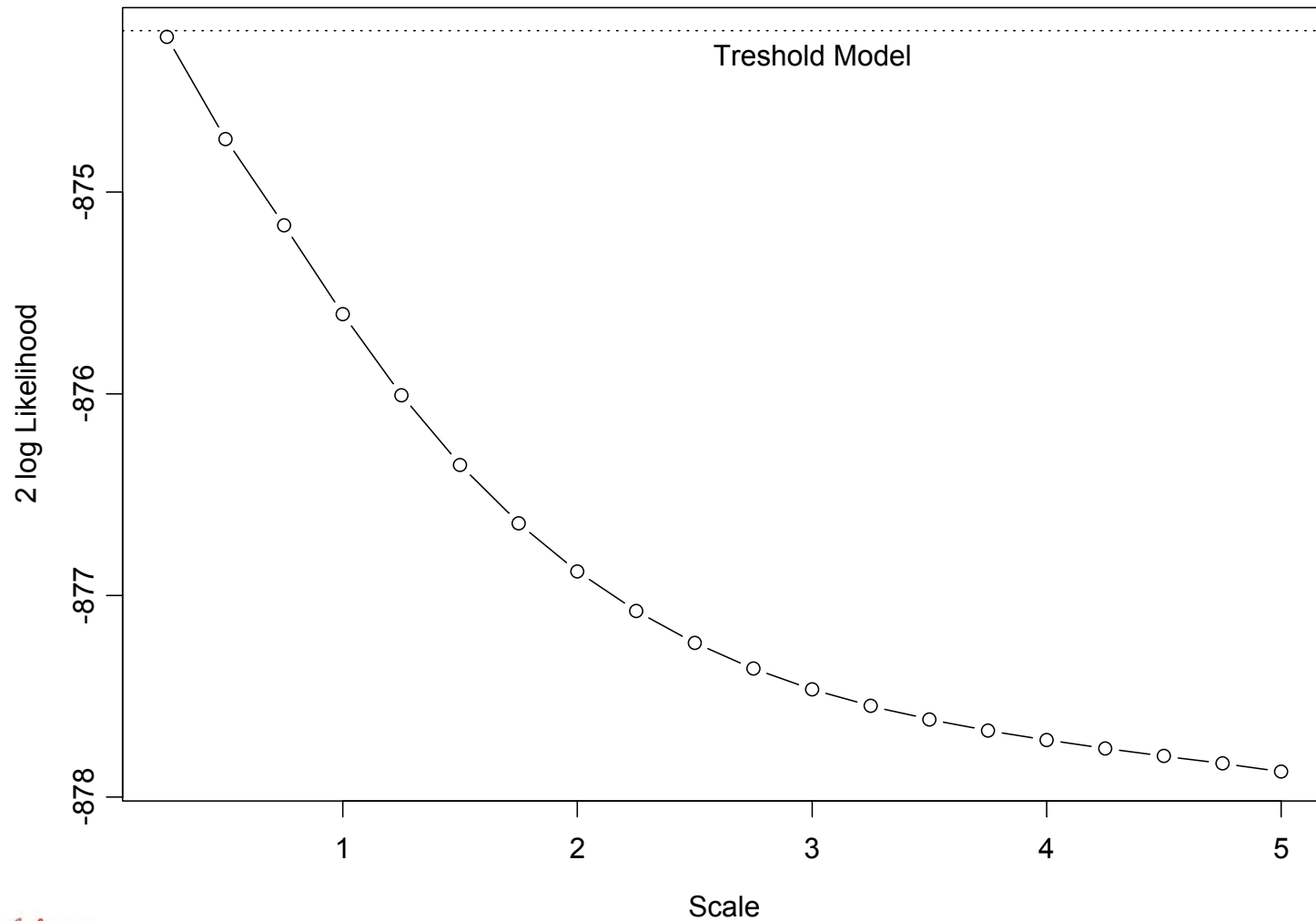
θ_1 : threshold for risk increase

θ_2 : log risk increase

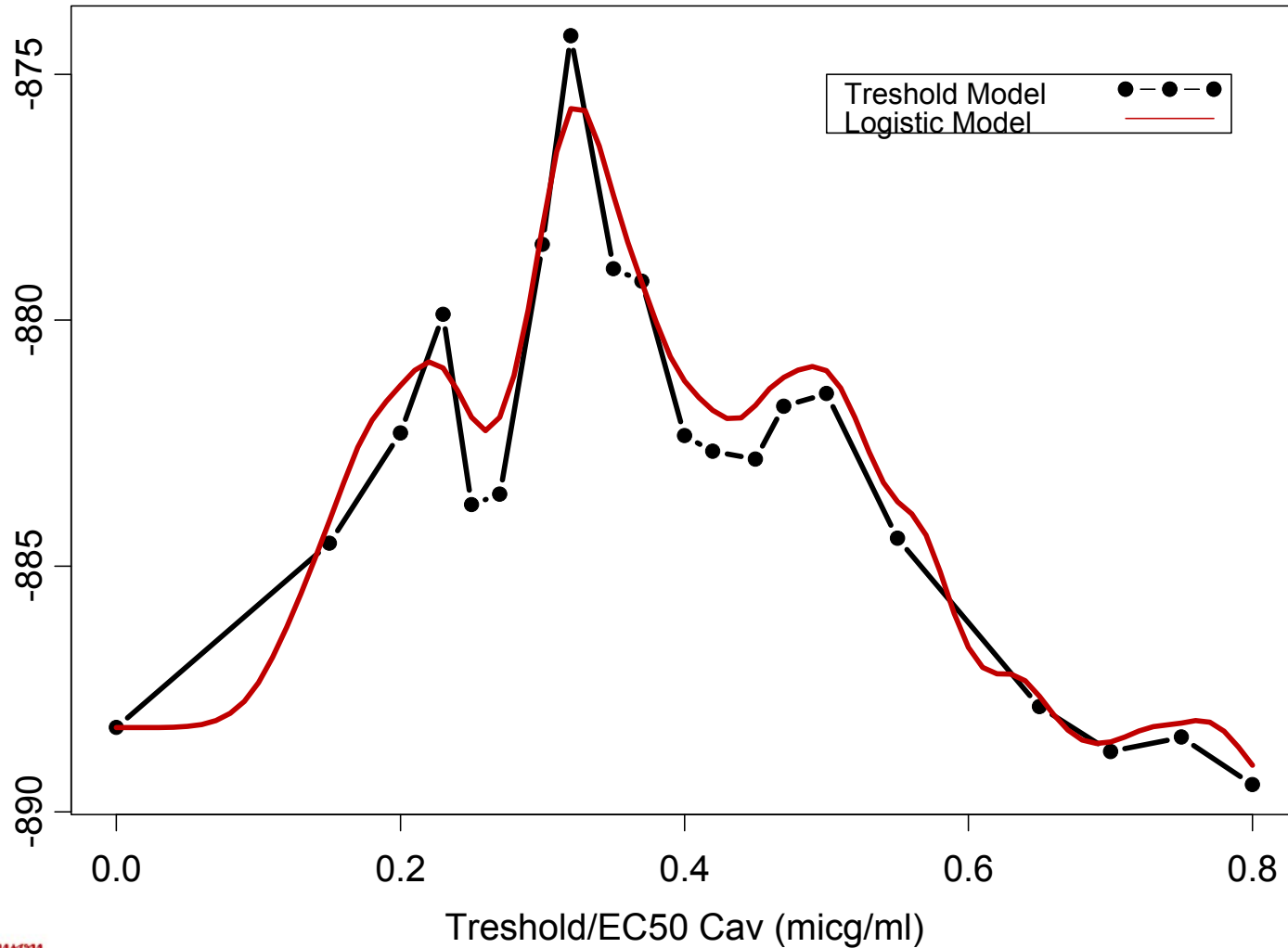
Functional form for various θ_3



Likelihood profile for θ_3



Threshold versus logistic model ($\theta_3=1$)



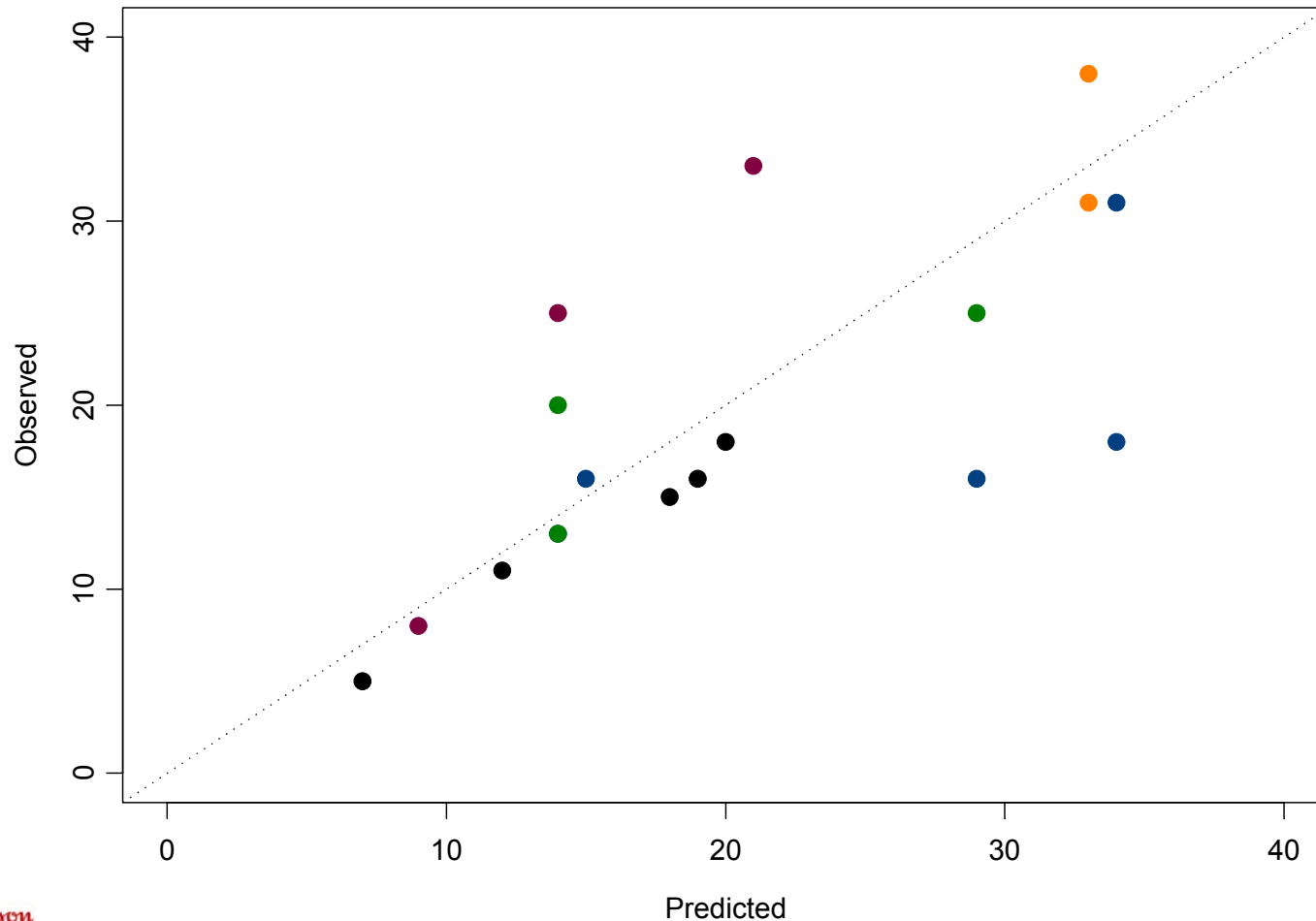
Implementation in SAS 8.2

```
id time cav event;  
1 1 12 0  
1 2 12 0  
1 3 12 0  
...  
1 60 12 0 = censored at Day 60  
2 1 65 0  
2 2 65 0  
2 3 65 1 = has AE at Day 3  
...
```

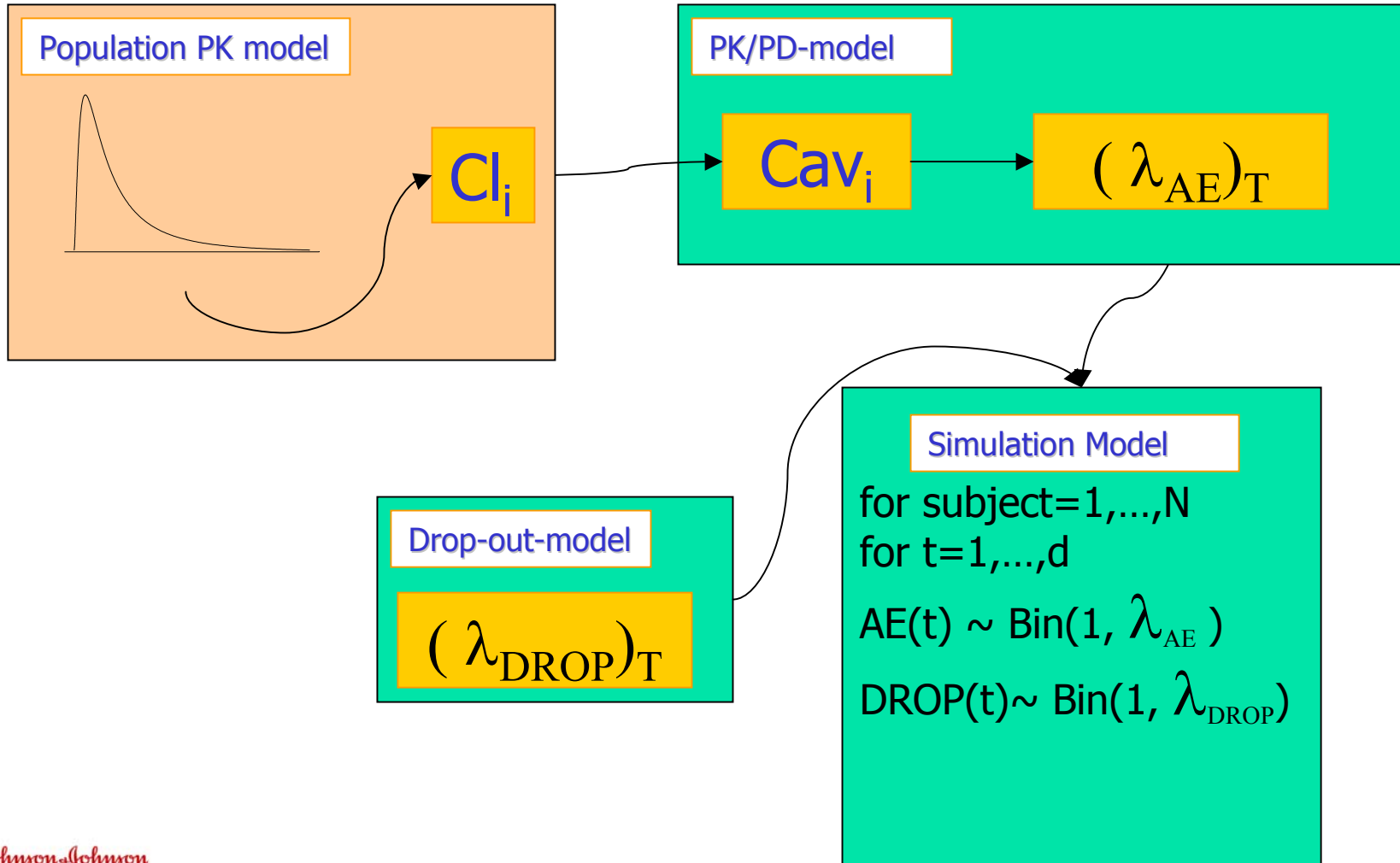
```
proc nlmixed;  
parms a=-4 b=-0.04 emax=1 ec50=3 scale=4;  
loghaz=a+b*time+emax/(1+exp(ec50-cav)/scale);  
p=exp(loghaz);  
model event~binomial(1,p);  
run;
```

Model Validation

AE incidence (%) in 5 independent clinical trials with drug X (total ± 2000 pat.). Each point is a different dose, each color a different trial

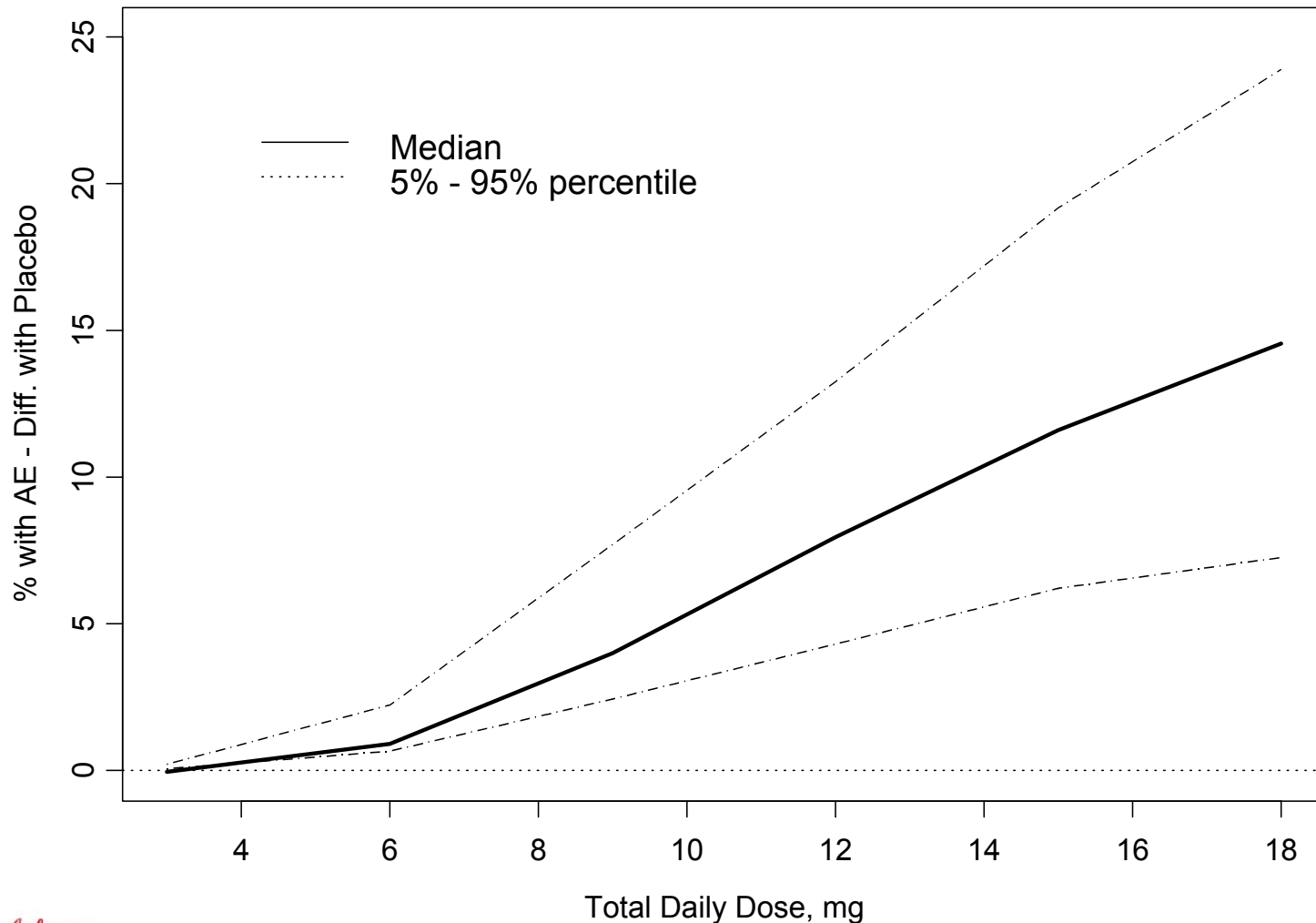


Clinical Trial Simulation for Drug Y



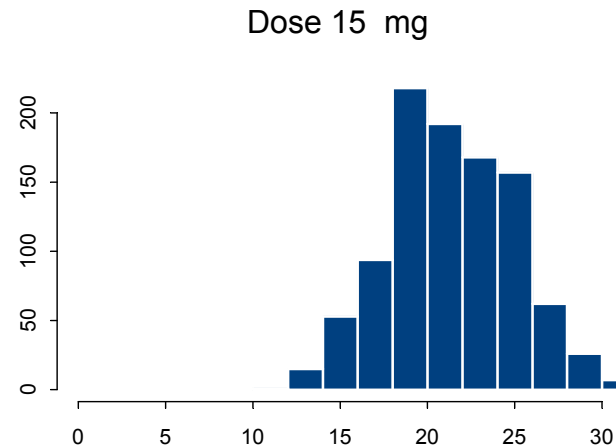
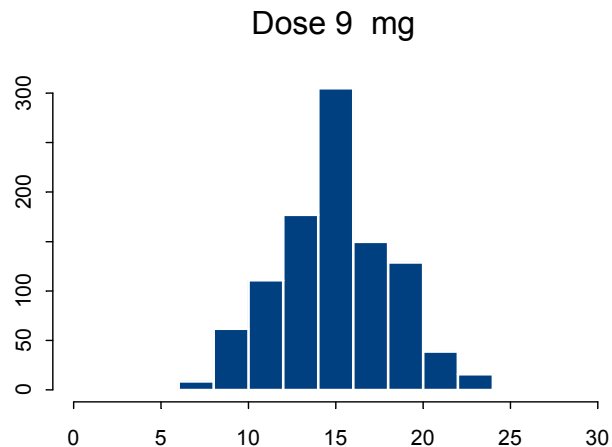
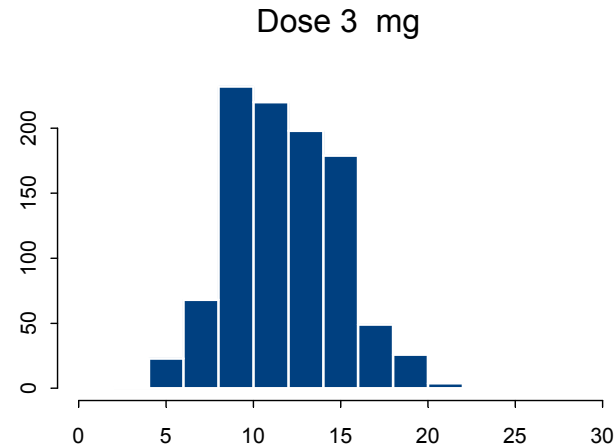
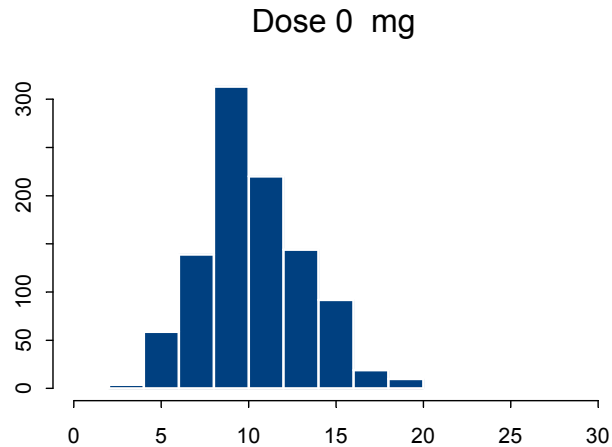
Incidence of AE in the population

Simulation of 2000 patients/dose, no drop-out, 8-week trial



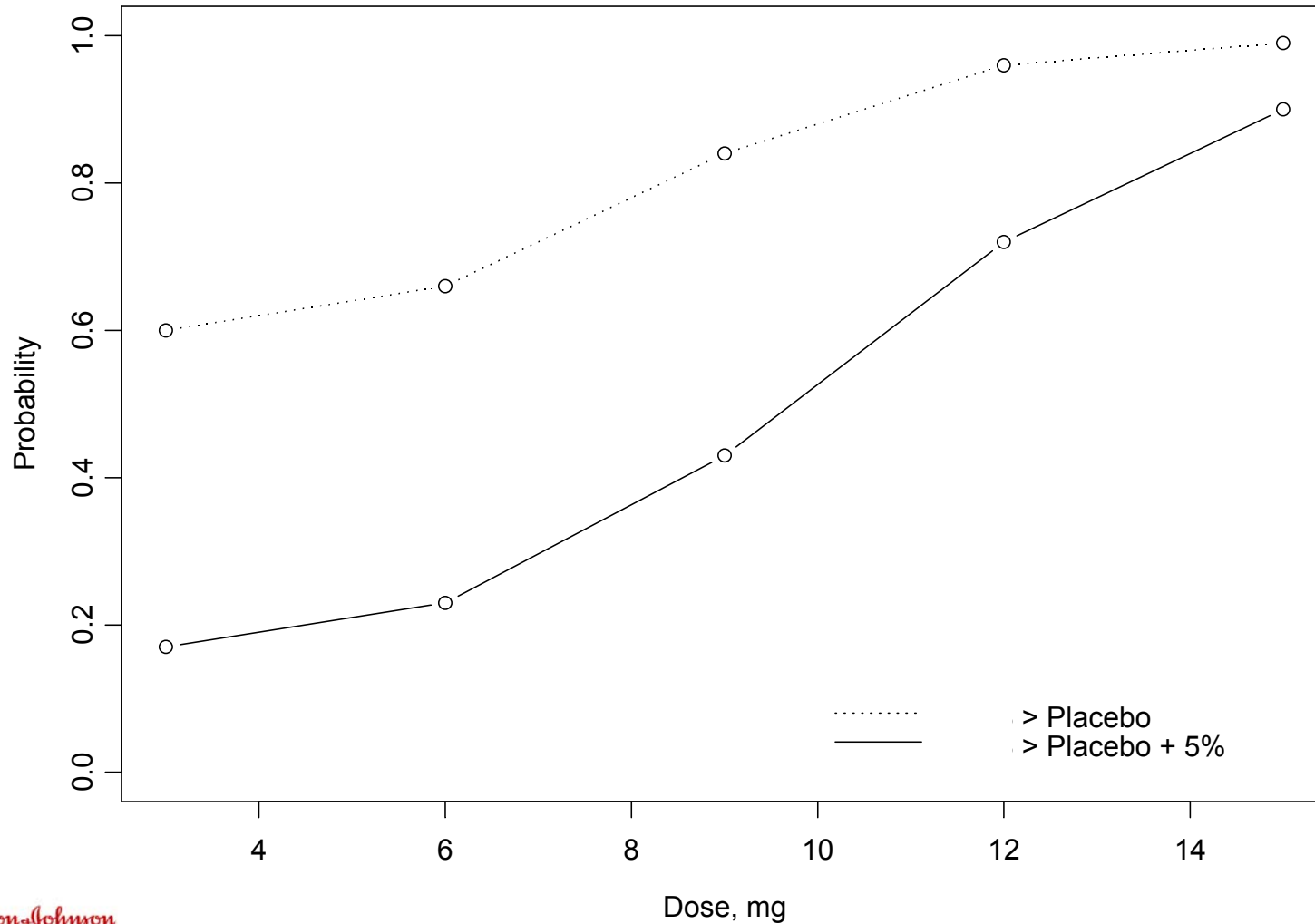
AE Incidence (%) in planned trials

Simulation of 1000 trials, 100 patients/dose, drop-out, 8-weeks



AE Incidence (%) in planned trials

Simulation of 1000 trials, 100 patients/dose, drop-out, 8-weeks



Conclusions

- ◆ Hazard modeling is a flexible tool to model the occurrence of side effects in clinical trials
- ◆ In the example presented here, the risk for a specific adverse event showed a steep (~on/off) relation with the patient-specific average steady state concentration
- ◆ Modeling allowed synthesis of data of a large number of studies of a marketed drug
- ◆ Clinical Trial Simulation allowed to study what can be expected in planned clinical trials, and optimize the dose range to study.