

# **R vs. SAS in model based drug development**

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**UseR!  
2009**

# Overview

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1. Introduction to model based drug development
2. An example
3. Current practice in the pharmaceutical industry
4. Planning how to model the development of a treatment
5. Coding and running the simulations: R vs. SAS
6. (Results of our simulation of a development program)
7. Summary

# Model based drug development

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- ◆ “Model based drug development ” a very fuzzy term. Can mean...
  - using a regression model to explore the dose-effect profile of a treatment (as opposed to comparing results for each dose vs. results from placebo group)
  - modelling the biological mechanism of the treatment– “pharmacometrics”
  - modelling a study or studies
    - simulating pseudo-subjects in a drug development program
      - » (early dose-finding phase IIa -> dose selection phase IIb -> pivotal final phase III)

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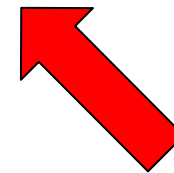
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This presentation

# An example

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- ◆ Challenging program: use model based drug development
  - model/simulate the working of the three phases, automating a mechanism for choosing doses at end of each phase
    - phase IIa -> phase IIb: choose highest safe dose
    - phase IIb -> phase III: choose safe dose with highest model-predicted efficacy
  - little data available about efficacy and safety in humans needing to be treated, so **under a number of possible scenarios**, evaluate
    - probability of finding viable dose
    - probability of finding the best dose (most efficacious dose that is safe)
  - “**scenario**”: profile of safety + profile for efficacy across candidate doses of treatment, including zero dose (placebo)

# Current practice in the pharma industry?

- ◆ Many large pharmaceutical companies talk of their model based drug development initiatives
  - in most cases this may describe work modelling the mechanism of action of the drug (simulating biology), not clear that it includes modelling a development program (simulating a chain of clinical trials)
- ◆ Business Week: “Pfizer, GlaxoSmithKline and others are using simulation...Novartis is far ahead of the rest of the industry” (again unclear what kind of simulation to which Business Week is referring)



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# Model based drug development: planning



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## ◆ Clear and detailed specification

- In Quintiles: specification is followed independently by originator and by quality-control (QC) statistician
  - two simulation programs produced
  - the two programs produce same or similar outputs
  - Originator used R
  - QC used SAS
- opportunity to compare R vs. SAS in this work
- since SAS and R cannot produce simulation of identical random variables, QC of the generation of data is best separate from QC of the processing of those variables
  - 1. QC the generation of random variables
  - 2. share the random data

# /Model based drug development: planning

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## ◆ /Clear and detailed specification

- facilitate using same programs for other treatments
  - self-contained functions in R for data generation, simulated analysis, dose selection
  - self-contained macros in SAS
- rules for selection of doses from a dose-effect curve
  - many simulations -> some unexpected outcomes – rules must cater for these
  - e. g. unsafe dose between two safe doses – assess the higher dose as unsafe?
- agree how to deal with simulations where development is discontinued mid-program
  - “empty” simulations passed to next stage of simulation?
  - “make up the numbers” by drawing at random from the simulations where the development was not stopped?

## R vs SAS

- ◆ **rv** package in R: nice compact generation of simulated data
  - its objects (=lists) not intended for use in modelling or other complex statistics

◆ Bulk of code: 972 lines (R) vs. 1323 lines

◆ Example

```
#simulate a placebo binary var and a group of active-group binary vars
ppe <- 0.15
pevents<- rvmnom(n=1, size=controls, prob=ppe)
pae <- max(.01, min(rvnorm(1, mean=activerate, sd=0.0001), 1))
aevents <- rvmnom(n=1, size=actives, prob=pae)
pnon <- controls - pevents
anon <- actives - aevents
```

R

```
do d=1 to maxdose;
    cohort=d;
    *** randomly perturb the fixed SAE rate for each simulation ***;
    rate_sp=max(.01, rand('NORMAL', saerate_dose(d), 0.0001));

    do pt=1 to ptN;
        subjid=d*1000+pt;
        dose=1;
        sae=rand('BERNULLI', rate_sp);
        if sae=0 then sae=2;
        output;
    end;

    rate_sp=max(.01, rand('NORMAL', saerate_dose(0), 0.0001));
    do pt=1 to ptN/2;
        subjid=pt;
        dose=0;
        sae=rand('BERNULLI', rate_sp);
        if sae=0 then sae=2;
        output;
    end;
end;
```

SAS

## R vs SAS

- ◆ **rv** package in R: nice compact generation of simulated data
  - its objects (=lists) not intended for use in modelling or other complex statistics

- ◆ Bulk of code: 972 lines (R) vs. 1323 lines

- ◆ Example

```
# simulate fishers exact test for placebo
(AE/death rate 15% vs. the active dose (rate specified)
for specified numbers of subjects
fishersim<-function(activate=0.15, controls, actives) {
  binfish <- sim.event(activate=activate, controls=controls, actives=actives)

  binfish$pnon<- controls - binfish$pevents
  binfish$anon<- actives - binfish$aevents

  rvfish<-rvmapply(function(x, y, a, b)
    fisher.test(
      matrix(c(x, a, y, b), nrow=2), alternative="1"
    ),
    x=binfish$pevents,
    a=binfish$pnon,
    y=binfish$aevents,
    b=binfish$anon
  )
  list(pvalue=rvfish$p.value)
}
```

R

```
proc freq data = Ila noprint ;
  tables dose*sae / nowarn fisher out=IlaSAEfreq(keep=scenario
sim cohort dose sae count where=(sae=1)) ;
  by scenario sim cohort;
  output out = IlaFish(keep=scenario sim cohort xpl_fish) exact ;
run;
```

SAS

# /Coding

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## R vs SAS

### ◆ Readability?

- R more compact, but its density may render it less readable for some users

# Coding and running

## R vs SAS

### ◆ Repeated processing of the same analysis required for simulation:

- R, “for (i in 1:n)”;
  - alternative: **rv** [=random variable] objects: compact simulation-based representation of random variables, useful for simple statistics, e.g.

```
» > setnsims(2500)
» > sim.binom <- rvsim(n=1, size=200, prob=0.20)an(sim.binom)
» > sim.binom
»   mean          sd 1% 2.5% 25% 50% 75% 97.5% 99% sims
» [1] 40          5.6 28 29 36 40 44 51 53 2500
» > sim.binom**2
»   mean          sd 1% 2.5% 25% 50% 75% 97.5% 99% sims
» [1] 1625        458 729 841 1296 1600 1936 2655 2916 2500
```

- SAS, BY statement – built-in economies in re-use of matrices



# Coding and running

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R vs SAS, elapsed time for simulations of drug development program,  
same machine used

- ◆ caveat #1: different programming styles and skill levels for each program language
- ◆ caveat #2: no particular attempt to improve time-efficiency of R code
- ◆ caveat #3: elapsed timings dependent on workload of machine being used

# Coding and running

---

R vs SAS, elapsed time for simulations of drug development program,  
same machine used

- ◆ 2500 simulations, no modelling involved
  - R: 10.65 minutes
  - SAS: 5 minutes

# Coding and running

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R vs SAS, elapsed time for simulations of drug development program,  
same machine used

- ◆ 2500 simulations, no modelling involved
  - R: 10.65 minutes
  - SAS: 5 minutes
  
- ◆ simulation involving 4 logistic regressions
  - 2500 simulations, 7 scenarios, (run at night)
    - SAS: 18 minutes
    - R: 7+ hours
  
  - logistic regression alone, (run during daytime)
    - R: 77.93 minutes for 2500 simulations of **single scenario**
    - SAS: 8.77 minutes for 2500 simulations of **7 scenarios!**

# Summary

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- ◆ Thorough QC of simulations is recommended when modelling a “chain” of clinical trials
    - requires careful specification
    - QC the generation of random variable data, which may not be matched across languages like SAS and R, separately from the processing of that data
  - ◆ Problem: how to simulate selection of dose where this is not based on quantitative rule, as in some dose-escalation studies?
  - ◆ tricky points in programming
    - “empty” simulations (where there was a safety issue or there was no efficacy detected, so development stopped for that simulation)
    - specify the structure of the simulation as well as the algorithms and analyses
      - » facilitate QC
      - » facilitate re-use of code for other the development of other treatments

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## ◆ R vs. SAS

- R compact, sometimes less readable
  - R's rv package nice for generating random variables, but of limited use for e.g. repeated simulations of ANCOVA or logistic regression
- SAS somewhat faster than R for simulations where no generalised linear modelling involved
- SAS considerably faster than R where generalised linear modelling is involved in simulations (by a factor of c.30)

# Extra slides

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# An example

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## ◆ Challenging drug development program

- mechanism of action of treatment not well understood
- data only from animals and healthy subjects
- broad options:
  - high risk, explore few doses
  - cautious approach, more doses
- outline of program:
  - small phase IIa to identify highest safe dose (as often in oncology)
  - phase IIb to identify dose for further development
  - phase III to provide enough evidence of safety and efficacy for regulatory approval
- **model based drug development** to help decide design of the phase IIb study

# Current practice in the pharma industry?

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## ◆ US, EU regulators

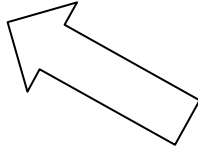
- Robert O'Neill, FDA: “Clinical scenario assessments are **much, much** more important than anything else that statisticians in drug development are doing at the moment, including adaptive design...”
- Robert Hemmings, EMEA: “I had a dream; 10 years from now there will be no drug development programs without clinical scenario simulations”



# /Current practice in the pharma industry?

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- ◆ Roche reported to work with Pharsight's "Trial simulator" 

Simulates trials, "point and click", not based on a language
- ◆ (For most adaptive clinical trials, simulation or complex integration is required to estimate power of the trial.) 

Addplan, Decimaker, East

User can add modules in R

# Results of our simulation of a development program

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- ◆ Difficult to perform simulation of small IIa study to find highest safe dose
  - in real development the decision about safety would be based on the totality of the clinical data: labs, adverse events, baseline scores, etc.
  - our simulation used only assumed proportion of SAEs/deaths



# /Results of our simulation of a development program

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Recall broad options for the sponsor:

◆ high risk, explore few doses

good success rate when no safety issues and all doses worked

◆ cautious approach, more doses

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strength was in scenarios where  
some doses were unsafe or  
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